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## Open Source Software Projects Needing Security Investments

David A. Wheeler, *Project Leader*

Samir Khakimov

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INSTITUTE FOR DEFENSE ANALYSES

IDA Document D-5459, v. 1.0

## **Open Source Software Projects Needing Security Investments**

David A. Wheeler, *Project Leader*

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## Executive Summary

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The Heartbleed vulnerability in the open source software (OSS) program OpenSSL was a serious vulnerability with widespread impact. It highlighted that some OSS programs are widely used and depended on and that vulnerabilities in them can have serious ramifications, and yet some OSS programs have not received the level of security analysis appropriate to their importance. Some OSS projects have many participants, perform in-depth security analyses, and produce software that is widely considered to be of high quality and to have strong security. However, other OSS projects have small teams that have limited time to do the tasks necessary for strong security.

The Linux Foundation (LF) Core Infrastructure Initiative (CII) is trying to identify OSS projects that need special focus/help for security so that it can best identify OSS projects needing investment. Similarly, the Department of Homeland Security Homeland Open Security Technology (DHS HOST) program's goal is to "help facilitate the continued adoption of open technology solutions (including OSS)...to improve system security...." They have asked the Institute for Defense Analyses to identify and collect metrics to help identify OSS projects that may especially need investment for security.

We have focused on gathering metrics automatically, especially those that suggest less active projects. We also estimated the program's exposure to attack. We developed a scoring system to heuristically combine these automatically gathered metrics with the exposure estimate. This scoring system identified especially plausible candidates. We took those candidates, examined their data further, and then identified a subset of candidates that we believe are especially concerning. We did not specifically examine the security of the projects themselves. The initial set of projects we examined was the set of software packages installed by Debian base (which are very widely used) combined with other packages that we or others identified as potentially concerning; we could easily add more projects to consider in the future.

This document provides information captured as part of our process to help identify open source software (OSS) projects that may need investment for security. It captures a brief literature search of ways to try to measure this and then describes the results we have captured so far.

This document is released under the Creative Commons Attribution 4.0 International (CC BY 4.0) license. The supporting software (in Python) for capturing data is released under the Massachusetts Institute of Technology (MIT) license. Some supporting data was

sourced from the Black Duck Open HUB (formerly Ohloh), a free online community resource for discovering, evaluating, tracking, and comparing open source code and projects.

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# 1. Introduction

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The Heartbleed vulnerability in the open source software (OSS) program OpenSSL was a serious vulnerability with widespread impact. Yet there are many ways that Heartbleed could have been detected before it was deployed [Wheeler2014h]. The Heartbleed vulnerability highlighted that the vulnerabilities in some widely used and depended-upon OSS programs can have serious ramifications, and yet they have not received the level of security analysis appropriate to their importance. Some OSS projects have many participants, perform in-depth security analyses, and produce software that is widely considered to be of high quality and to have strong security. However, other OSS projects have small teams that have limited time to do the tasks necessary for strong security (e.g., the OpenSSL project before Heartbleed).

The Linux Foundation (LF) Core Infrastructure Initiative (CII) was established to “fund open source projects that are in the critical path for core computing functions [and] are experiencing under-investment.”<sup>1</sup> The LF CII will make final decisions on what it will invest in, but it has asked for help in identifying appropriate metrics and their values.

The HOST program’s goal, as stated in the statement of work between the Institute for Defense Analyses (IDA) and the Georgia Tech Research Institute (GTRI), is to “help facilitate the continued adoption of open technology solutions (including OSS) within federal, state, and municipal public sector [information technology] IT environments in order to improve system security....” The program has asked IDA to “provide continued subject matter expertise on in-depth research, studies, and analysis on the research domain” for HOST. The HOST project is in turn funded by the Department of Homeland Security (DHS). This is in support of securing the nation—Executive Order 13636 states that, “It is the policy of the United States to enhance the security and resilience of the Nation’s critical infrastructure and to maintain a cyber environment that encourages efficiency, innovation, and economic prosperity while promoting safety, security, business confidentiality, privacy, and civil liberties” [Obama2013]. Thus, identifying OSS projects that need help is also in the interest of the HOST project.

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<sup>1</sup> <http://www.linuxfoundation.org/programs/core-infrastructure-initiative>

Other organizations and projects might be interested in this work as well. For example:

- The “Snowdrift coop” (at <https://snowdrift.coop/>) was established in late 2014 to create “a matching patronage system funding freely-licensed works” and in the future might be interested in funding work to improve the security of OSS projects.
- The “European Parliament has approved funding for several projects related to Free Software and privacy. In the EU budget for 2015, which the European Parliament adopted on December 17, the Parliamentarians have allocated up to one million Euro for a project to audit Free Software programs in use at the Commission and the Parliament in order to identify and fix security vulnerabilities,” <https://fsfe.org/news/2014/news-20141219-01.en.html>.
- The “Google Application Security Patch Reward Program” rewards proactive security improvements in selected open-source projects. See <https://www.google.com/about/appsecurity/patch-rewards/>.

The Linux Foundation and HOST have asked us to identify and collect metrics to help identify OSS projects that may especially need investment for security.

The goal of this work was to perform a relatively quick reaction study to gather data to help make reasonable decisions in a short time. Further work might identify significantly improved measures and additional projects to be examined, but it was judged to be better to do a quick study with limited time (and document it) than to simply guess or to spend a long time on a more comprehensive study.

We began this work by surveying past and current efforts to identify relevant OSS project metrics (summarized in this document). We also identified various ways to identify which OSS projects might be considered especially important. We then selected an automated approach for gathering relevant metrics for a candidate set of important OSS projects, created a prototype, and then refined the prototype after examining its early results. During this process we participated in a 2015 London conference, where we shared our early results and received helpful feedback (including references to better data sources).

We have focused on metrics that we can gather automatically that suggest less active projects. We also estimated the program’s exposure to attack. We have developed a scoring system to heuristically combine these automatically gathered metrics with our estimate of attack exposure. These heuristics identified especially plausible candidates. We then examined those candidates further and identified a subset that we believe are especially concerning. The initial set of projects we examined was the set of software

packages<sup>2</sup> installed by Debian base, to which we added packages that we or others identified as potentially concerning; we could easily add more projects to consider in the future.

This document uses the term “open source software” (OSS) for software that can be studied, used for any purpose, modified, and redistributed (modified or not). Other terms for such software including “Free software” (note the capital letter) and “Free/libre/open source software” (FLOSS). See the Open Source Definition [OSI] and the Free Software Definition [FSF] for details. In some cases the users of these different terms emphasize different motivations and purposes, but since we are simply focused on the software *resulting* from these efforts (instead of the motivations for development), we will ignore those distinctions in this paper. We use “proprietary software” and “closed software” as antonyms for OSS. Note that “in almost all cases, OSS meets the definition of ‘commercial computer software’” under U.S. law [DoD2009] and that many OSS programs are co-developed and supported by commercial companies.

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Chapter 2 lists past work, identifying relevant ways to measure OSS projects. Chapter 3 is a list of especially promising metrics, based on Chapter 2, for measuring OSS projects’ need for security investment. Chapter 4 identifies important OSS projects that are widely used yet might need investment. We are developing software to capture and combine this data into a separate spreadsheet for developing recommendations based on this document.

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<sup>2</sup> A software package is a unit of software that can be easily installed, updated, and uninstalled using a tool called a “package manager.” Packages include dependency information, making it possible to automatically install other packages a given package depends on. Common package formats include the .deb format (used by Debian and Ubuntu) and the .rpm format (used by Red Hat Enterprise Linux and Fedora).



## **2. Some Past and Current Efforts to Identify Relevant OSS Project Metrics**

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This chapter describes a brief survey (literature search) of some past efforts to identify relevant metrics of OSS projects. Our goal was to help identify metrics that might help identify projects needing investment.

Measuring the security of software is a notoriously difficult and an essentially unsolved problem. Ideally we would identify metrics that directly determine whether or not an OSS project is producing secure software. However, since perfect metrics are not available, we are instead interested in metrics that provide some evidence that a project's product is more or less likely to be secure. Some product measures, for example, may suggest that the software has fewer security defects, or at least fewer defects in general. Other metrics examine the OSS project (including its processes) and may suggest that an OSS project is in trouble (e.g., it is relatively inactive, has few active contributors, or that much development was done long ago (when fewer developers knew how to develop secure software)). For example, it is often noted that before Heartbleed, OpenSSL had relatively few developers and that many bug reports languished without response for long periods of time. These indicators may suggest that a project needs investment to make its software adequately secure.

Sources include surveys of OSS, existing evaluation processes for evaluating OSS, surveys of quality or security metrics (e.g., [Shaikh2009]), and organizations that track OSS metrics.

There was not time to do a complete survey, but we believe it is better to do a brief survey (and document it) than ignore the large set of materials available. These materials are probably not equally useful or credible; the goal was simply to survey various options to reduce the risk of overlooking especially useful sources of information. Some odd or improbable approaches might suggest a new and useful approach.

### **A. OSS Metrics Data Sources**

It is much easier to get data from organizations that measure and curate it than to try to extract it for each program separately. There is also the hope that such organizations will try to select useful measures. Black Duck Open Hub (formerly Ohloh), in particular, provides relatively current data for many programs in an easily obtained form.

## 1. Black Duck Open Hub

Black Duck Open Hub, formerly Ohloh, maintains an active set of metrics data for a variety of OSS projects at <https://www.openhub.net>, along with a nice user interface (UI) for viewing them.

Looking at a sample project entry, such as Firefox, helps give a sense of what is recorded for a project. The entry for Firefox (<https://www.openhub.net/p/firefox>) in December 2014 reports:

“In a Nutshell, Mozilla Firefox...

- has had 223,200 commits made by 3,187 contributors representing 12,554,058 lines of code.
- is mostly written in C++ with a low number of source code comments [as a percentage compared to other programs in the same programming language].
- has a well established, mature codebase maintained by a very large development team with increasing year-over-year (Y-O-Y) commits.
- took an estimated 3,920 years of effort (COCOMO model) starting with its first commit in April, 2002, and ending with its most recent commit 26 days ago.”

It reports, for both 30-day and 12-month periods, the number of commits and the number of contributors (including a separate number for new contributors). For the 12-month period it also reports the change from the previous 12-month period. It includes user ratings.

It also provides “quick reference” information (such as the organization name), some of which can also indicate the health of a project:

- Link(s) for Homepage, Documentation, Download, Forums, Issue Trackers, and Code: Where present, these are signs of an active project.
- Licenses: OSI- and FSF-approved licenses, especially if they are common, are a good sign because unusual licenses can inhibit contribution.

This is a well-maintained site with programmatic interfaces that make it easy to access the data they collect. The programmatic interfaces in some cases have only general statements (e.g., “mature codebase” or “very large development team”) instead of specific numbers, but these general statements can still be valuable.

A (gratis) key must be acquired for programmatic queries, and the website states that queries are limited to 1,000 queries/day for each key (although this might not be enforced). We cached results to avoid creating a nuisance. However, they also impose other conditions, so we arranged a special exception with Black Duck for use in this project.



## **2. OSS Repository Statistics (GitHub, SourceForge, git, etc.)**

Many OSS projects are hosted on a relatively few number of hosting sites that can also report a variety of statistics. If the repository (also called a “repo”) directly provides that data, then the data is especially easy to get for such projects.

GitHub provides a variety of statistics and charts, particularly ones focused on project activity. Selecting “pulse” on a project’s project site accesses reports for a selected time period on the number of (direct) authors, commits, files, and number of additions and deletions (counted by lines). You can also select “graphs” to see a variety of graphs. The issue tracker can report the number of open and closed issues. More information is available via <https://developer.github.com/v3/repos/statistics/> for programs that need this data.

SourceForge has switched to the OSS Allura software for repository management. Its command “Tickets/View Stats” reports a variety of statistics, including number of tickets (total, open, closed), number of new tickets over various periods (7 days, 14 days, 30 days), number of comments on tickets, and number of new comments on tickets over a given period. They also support a Representational State Transfer (REST) application programming interface (API) for obtaining this information for programs (most of its data is returned in JSON format); more information is at <https://sourceforge.net/p/forgedocumentation/Allura%20API/>.

Distributed version control software, including git, includes a significant amount of metadata about commits because the project history is downloaded. Tools such as gitstats (<http://gitstats.sourceforge.net/>) can be used to quickly analyze this data and report additional information. Gitstats, for example, will report:

- General statistics: total files, lines, commits, authors
- Activity: commits by hour of day, day of week, hour of week, month of year, year and month, and year
- Authors: list of authors (name, commits (%), first commit date, last commit date, age), author of month, author of year
- Files: file count by date, extensions
- Lines: Lines of Code by date.

## **3. Linux Distribution Repositories**

Most Linux distributions (such as Debian, Ubuntu, Fedora, and Red Hat Enterprise Linux) use package managers to install (and uninstall) packages. These packages include metadata with important information, such as the software name, dependencies, and URL of the originating project. Additionally, if a package is installed in a distribution’s base or a widely used group/task, it is likely to be widely used. Some distributions (such as Fedora)

work to split up projects so that if software from another project is reused, the projects are kept separate (so that security updates will properly update everything); this can help reveal important projects that might otherwise be hidden inside larger projects.

#### **4. FLOSSmole**

FLOSSmole at <http://flossmole.org/> performs “collaborative collection and analysis of free/libre/open source project data” (per its front page). It is related to FLOSShub, a “portal for free/libre and open source software (FLOSS) research resources and discussion.”

It appears semi-active, with some datasets and various reports dated 2014. However, some data is only available through 2013. One challenge for them is that repositories are increasingly providing this information directly.

#### **5. FLOSSMetrics Project**

FLOSSMetrics stands for “Free/Libre Open Source Software Metrics” and is at <http://www.flossmetrics.org/>.

The main objective of FLOSSMETRICS is, per its website, “to construct, publish and analyse a large scale database with information and metrics about libre software development coming from several thousands of software projects, using existing methodologies, and tools already developed.”

It records various data for a variety of projects, for example:

- How many bugs are reported
- The average time it takes to fix a bug in a project’s lifetime.

There was a “final report” in 2010 for this European project, and no obvious activity since then. Its database of projects at <http://melquiades.flossmetrics.org/projects> seems to have had little activity since 2010. Thus, this is likely to be no longer active. Active similar projects include FLOSSmole and Black Duck Open Hub.

#### **6. FLOSS Community Metrics Meeting**

The “FLOSS Community Metrics” meeting is a conference of those interested in collecting and analyzing OSS metrics, sponsored by Bitergia. Its website is at <http://flosscommunitymetrics.org/>; they had a conference in July 2014 and another is expected in 2015. The 2014 conference had several presentations on measuring OSS quality, which are summarized below (clicking on the “slides” link on its website provides the slides described below).

Roberto Galoppini’s presentation, “You’re not entitled to your opinion about open source software!” proposed the following simple-to-collect metrics:

- Code Maturity [<1 year, 1-3 years, > 3 years]
- Code stability (unstable, stable but old, stable and maintained)
- Project popularity (unknown, small but growing, well known)
- Case study availability
- Books availability
- Community management style
- Team size [1-5 members, 5–10 members, > 10 members]—can be found by analyzing commits
- Commercial support
- Training
- Documentation
- QA Process [n/a, existing but not supported by tools, supported by tools]
- QA tools [n/a, existing but not much used, very active use of tools]
- Bugs reactivity [poor, formalized but not reactive, formalized and reactive]
- Source [to be compiled, binaries available, virtual appliance available]
- Red Hat/Solaris/Windows
- Amount of comments [none, poorly commented, well commented]
- Computer language used [more than 3 languages used, 1 language primarily, 1 unique language]
- Code modularity [not modular, modular, available tools to create extensions]
- License
- Modifiability [no way to propose modification, tools to access and modify code available but the process is not well defined, tools and procedures to propose modifications available.]
- Roadmap [n/a, no detailed roadmap available, detailed roadmap available]
- Sponsor.

James Faulkner (Liferay community manager) presented “Metrics are fun, but which ones really matter?”; this presentation lists various metrics and identified those he thought were “more interesting”:

- Time of bug report to fix
- Time from forum question to answer

- Number of ignored contributions
- Time from contribution to insertion in codebase.

He also lists “basic” metrics such as number of contributions, number of commits/lines, number of authors, number of bug reports, number of forum posts/answers, number of downloads, number of ignored messages, and number of open tickets/code reviews.

The conference proceedings also noted vizGrimoire, an OSS toolset and framework to analyze and visualize data about software development, available at <http://vizgrimoire.bitergia.org/> promoted by Bitergia.

## 7. Rodriguez Survey of Software Data Repositories

Rodriguez et al’s “On Software Engineering Repositories and their Open Problems” describes various sources of data about software [Rodriguez2012]. They identified the following set:

- FLOSSMole: <http://flossmole.org/>
- FLOSSMetrics: <http://flossmetrics.org/>
- PROMISE (PRedictOr Models In Software Engineering): <http://promisedata.org/>
- Qualitas Corpus (QC): <http://qualitascorpus.com/>
- Sourcerer Project: <http://sourcerer.ics.uci.edu/>
- Ultimate Debian Database (UDD): <http://udd.debian.org/>
- Bug Prediction Dataset (BPD): <http://bug.inf.usi.ch/>
- International Software Benchmarking Standards Group (ISBSG): <http://www.isbsg.org/>
- Eclipse Bug Data (EBD) <http://www.st.cs.uni-saarland.de/softevo/bug-data/eclipse/Software-artifact>
- Infrastructure Repository (SIR): <http://sir.unl.edu/>
- Ohloh: <http://www.ohloh.net/>
- SourceForge Research Data Archive (SRDA): <http://zerlot.cse.nd.edu/>
- Helix Data Set: <http://www.ict.swin.edu.au/research/projects/helix/>
- Tukutuku: <http://www.metriq.biz/tukutuku/>.

## 8. PROMISE

PROMISE stands for “PRedictOr Models In Software Engineering”; its main website is <http://promisedata.org/> but its dataset is at <http://openscience.us/repo/>. This is a collection of data “real world software engineering projects...whatever data is available.” This is useful for tasks such as research into predictive metrics. However, it is a collection of *available* real-world data, not primarily *current* real-world data, and by itself, it does not identify metrics that are necessarily relevant (just what it can provide).

The PROMISE 2014 (“PROMISE ’14”) conference included various presentations, including a keynote by Audris Mockus (at <http://mockus.org/papers/promise1.pdf>) on the problems of acquiring data and prediction.

## B. Methods for Evaluating OSS Projects

A number of complete processes are specifically for evaluating OSS (as software, a project, or both)—typically for a particular purpose, and not their security per se, but some of their approaches may also be useful for our purposes.

Wikipedia includes a comparison of a few of these processes for evaluating OSS.<sup>3</sup>

For our purposes, a key attribute is whether or not the OSS evaluation process supports comparison between different OSS programs. Since there are limits to how much can be invested, it is important to be able to determine which projects most need investment; this means that there must be a way to compare the OSS projects. The Open Source Maturity Model (OSMM) from Capgemini and Qualification and Selection of Open Source software (QSOS), for example, do support comparison.

The Open Business Readiness Rating (OpenBRR) project, announced in 2005, did not create a community. Its website <http://www.openbrr.org/> claims to be preparing an update; since that has not yet occurred, we do not consider it here.

Some documents and books, such as [Fogel2013], provide guidance on how to develop open source software. These could be used to evaluate OSS projects by determining how well the guidance is followed. We did not pursue this due to lack of time.

### 1. Stol and Babar

There are so many processes for evaluating OSS that Stol and Babar have published a framework comparing them. <http://staff.lero.ie/stol/files/2011/12/OSS2010.pdf>. One complication is that several methods are all named “Open Source Maturity Model.”

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<sup>3</sup> [http://en.wikipedia.org/w/index.php?title=Open-source\\_software\\_assessment\\_methodologies&oldid=579922098](http://en.wikipedia.org/w/index.php?title=Open-source_software_assessment_methodologies&oldid=579922098)

## **2. QualiPSo OpenSource Maturity Model (OMM)**

The QualiPSo OpenSource Maturity Model (OMM) is a methodology for assessing Free/Libre Open Source Software (FLOSS) and more specifically the FLOSS development process. This methodology was released in 2008 and is released under the Creative Commons license.

The summary, at [http://en.wikipedia.org/wiki/OpenSource\\_Maturity\\_Model](http://en.wikipedia.org/wiki/OpenSource_Maturity_Model), defines three maturity levels: basic, intermediate, and advanced.

The basic-level requirements are:

- PDOC – Product Documentation
- STD – Use of Established and Widespread Standards
- QTP – Quality of Test Plan
- LCS – Licenses
- ENV – Technical Environment
- DFCT – Number of Commits and Bug Reports
- MST – Maintainability and Stability
- CM – Configuration Management
- PP1 – Project Planning Part 1
- REQM – Requirements Management
- RDMP1 – Availability and Use of a (product) roadmap.

The intermediate-level requirements are:

- RDMP2 – Availability and Use of a (product) roadmap
- STK – Relationship between Stakeholders
- PP2 – Project Planning Part 2
- PMC – Project Monitoring and Control
- TST1 – Test Part 1
- DSN1 – Design Part 1
- PPQA – Process and Product Quality Assurance.

The advanced-level requirements are:

- PI – Product Integration
- RSKM – Risk Management
- TST2 – Test Part 2

- DSN2 – Design 2
- RASM – Results of third party assessment
- REP – Reputation
- CONT – Contribution to FLOSS Product from SW Companies.

Unfortunately, we have had trouble accessing <http://www.qualipso.org/> for more information.

### 3. QSOS

Qualification and Selection of Open Source Software (QSOS) dates from 2004; QSOS 2.0 was released June 2013. It is a “free project aiming to mutualize and capitalize technological watch on open source components and projects”; its main page is <http://www.qsos.org>. The method is distributed under the GNU Free Documentation License. Its goal is to provide a “formal process to evaluate, compare and select open source solutions.”

QSOS is a Drakkr project (<http://www.drakkr.org>); others include FLOSC (Free/Libre Open Source Complexity), a project providing method and tools to evaluate the intrinsic complexity of open source components. Some information is available on GitHub <https://github.com/drakkr/drakkr>.

Unfortunately the site <http://master.o3s.qsos.org/> was down when we started this work, and when it finally resurfaced it was in French, so we have not delved into it further at this time.

### 4. SQO-OSS / Spinellis, et al

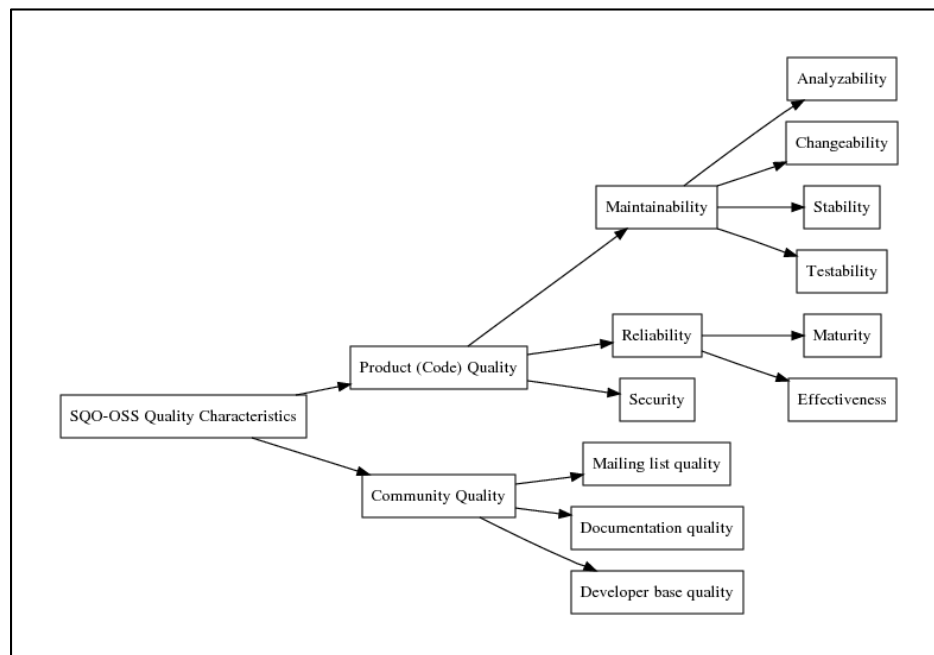
“Evaluating the Quality of Open Source Software” [Spinellis2009] presents “motivating examples, tools, and techniques that can be used to evaluate the quality of open source... software.” It includes a “technical and research overview of [Software Quality Observatory for Open Source Software (SQO-OSS)], a cooperative research effort aiming to establish a software quality observatory for open source software.”

The paper notes the following metrics:

- Use of various scanning tools including PMD (Java scanner), FindBugs (for Java), Checkstyle (Java style checker), Sonar, ESX (for C++), and Scan by Coverity
- Use of metric suites such as Ohloh and Sourcekibitzer (the latter is for Java)
- Adherence to claimed coding style as a proxy for quality (They formatted FreeBSD code using indent and computed the number of lines that changed.)

- Mean developer engagement (MDE), the average percentage of active developers who work on a project each week (Developers who stay inactive are eventually considered no longer part of the total.)
- Cross-language metric tool, which collects metrics such as number of public attributes, number of children, etc.
- Metric for developer contributions. This adds measures for not just lines of code, but also for bug closing, documentation files, updating a wiki page, etc. Some measures are difficult to measure using only automated tools (e.g., “participate in a flamewar”).

It notes the SQO-OSS quality model shown in Figure 1.



**Figure 1. The SQO-OSS Quality Model**

And, it uses the metrics shown in Table 1 to estimate them.

**Table 1. Metrics used by the SQO-OSS Quality Mode**

Attribute	Metric
Analyzability	Cyclomatic number
	Number of statements
	Comments frequency
	Average size of statements
	Weighted methods per class ( WMC)



Attribute	Metric
	Number of base classes
	Class comments frequency
Changeability	Average size of statements
	Vocabulary frequency
	Number of unconditional jumps
	Number of nested levels
	Coupling between objects (CBO)
	Lack of cohesion (LCOM)
	Depth of inheritance tree (DIT)
Stability	Number of unconditional jumps
	Number of entry nodes
	Number of exit nodes
	Directly called components
	Number of children (NOC)
	Coupling between objects (CBO)
	Depth of inheritance tree (DIT)
Testability	Number of exits of conditional structs
	Cyclomatic number
	Number of nested levels
	Number of unconditional jumps
	Response for a class (RFC)
	Average cyclomatic complexity per method
	Number of children (NOC)
Maturity	Number of open critical bugs in the last 6 months
	Number of open bugs in the last 6 months
Effectiveness	Number of critical bugs fixed in the last 6 months
	Number of bugs fixed in the last 6 months
Security	Null dereferences
	Undefined values
Mailing list	Number of unique subscribers
	Number of messages in user/support list per month
	Number of messages in developers list per month
	Average thread depth
Documentation	Available documentation documents
	Update frequency
Developer base	Rate of developer intake

Attribute	Metric
	Rate of developer turnover
	Growth in active developers
	Quality of individual developers

The authors rank various projects by comparing values to ideal values, (e.g., the ideal candidate for the Excellent Analyzability quality attribute should have a McCabe Cyclomatic number equal to 4, an average function's number of statements equal to 10, a comments frequency equal to 0.5, and average "size of statements" equal to 2).

## 5. Ghapanchi's Taxonomy for Measuring OSS Project Success

Ghapanchi et al's "A taxonomy for measuring the success of open source software projects" [Ghapanchi2011] used a literature survey focused on measuring OSS project success. After identifying 154 publications in their initial set, they narrowed it down to 45 publications and categorized them into meaningful clusters.

They identified six broad areas that can lead to success:

1. Product quality. Different researchers have proposed many different measures for product quality (leading to product success). They report that [Crowston2003] is among the most cited; this is a content analysis of an online focus group that reported seven main themes for OSS success: user, product, process, developers, use, recognition, and influence.
2. Project performance: These combine efficiency and effectiveness.
3. Project effectiveness: These attempt to measure "getting the right things done."
4. Project efficiency: These determine the extent to which a project uses its resources to generate outcomes, typically using Data Envelopment Analysis (DEA). The goal is to determine Output/Input. Examples of input/output pairs are {number of developers, bug submitters}/{KiB per download, number of downloads, project rank}; {number of downloads, number of years}/{product size in bytes, number of code lines}; {product size (bytes), development status}/{number of developers, product age}.
5. Project activity: This is "frequently regarded as one of the pillars of OSS project success." Examples include how frequently defects are fixed, new releases of the software are posted, or support requests are answered, often over a period of time.
6. User interest: The ability of an OSS project to attract community members to adopt the software (its popularity).

They note that “success” can be measured for both the product and the project, so they then map these areas to these kinds of success. User interest affects both product success and project success. They map product quality to primarily product success, and map project activity, project efficiency, and project effectiveness to project success. Project effectiveness and project efficiency themselves affect project performance.

Based on this survey, they provide a “practical list of OSS success metrics,” see Table 2.

**Table 2. Practical List of OSS Metrics**

ASPECT	USEFUL MEASURES ACCORDING TO [GHAPANCHI2011]
User interest	Traffic on the project Web site, downloads of the code, number of developers who have joined the project team, and the number of people who have registered on the project mailing list to receive announcements such as new release regarding a project
Project activity	The number of software releases, number of patches, number of source code lines, number of code commits
Project effectiveness	The percentage of task completion (bug fix, feature request, and support request), number of developers the project has attracted, number of work weeks spent on the project
Project efficiency	Using a DEA model with one or some input indicators (e.g., number of developers, number of bug submitters, number of years, product size (bytes), development status) and one or some output indicators (e.g., kilobytes per download, number of download, project rank, product size in bytes, number of code lines)
Product quality	Code quality, documentation quality, understandability, consistency, maintainability, program efficiency, testability, completeness, conciseness, usability, portability, functionality, reliability, structuredness, meeting the requirements, ease of use, user friendliness

Source: Ghapanchi 2011

They make the interesting observation that the kinds of data available for OSS are typically different than for proprietary software. “Traditional [closed source] software development success models frequently focus on success indicators such as system quality, use, user satisfaction and organizational impacts [that are] more related to the ‘use environment’ of the software, while studies on OSS success tend to look more at the ‘development environment’... [in traditional models] ‘development environment is not publicly available but the ‘use environment is less difficult to study, while in OSS the ‘development environment’ is publicly visible but the ‘use environment’ is hard to study or even to identify.”

One challenge for us is that we are interested primarily in projects that are *successful* in terms of widespread adoption and satisfaction of functional requirements, yet have

serious *vulnerabilities*. These measures do not necessarily directly relate to our question, but they can perhaps help suggest projects that are not adequately active.

## 6. Wheeler OSS Evaluation Model

David A. Wheeler (also an author of this paper) previously described a general process for evaluating open source software in “How to Evaluate Open Source Software/Free Software (OSS/FS) Programs.” [Wheeler2011e].

This process is based on four steps: identify candidates, read existing reviews, compare the leading programs’ basic attributes to your needs, and analyze the top candidates in more depth. This set of *identify*, *read reviews*, *compare*, and *analyze* can be abbreviated as “IRCA.” Important attributes to consider include functionality, cost, market share, support, maintenance, reliability, performance, scalability, useability, security, flexibility/customizability, interoperability, and legal/license issues.

The section on security mentions the following metrics:

- Coverity scan results, including the rung achieved, number of defects, and defect density
- Fortify scan results (similar)
- Common criteria evaluation (These typically evaluate entire systems (e.g., entire operating systems), instead of focusing on specific projects that support a particular portion of an operating system, and thus do not provide the kinds of measures desired for this task.)
- Reports of (many) vulnerabilities that are “unforgiveable” (MITRE identifies criteria for identifying vulnerabilities that are especially easy to find, and thus “unforgiveable” [Christey2007].)
- Whether or not at least one external organization is known to have reviewed or be reviewing the software. However, some organizations that review software (such as OpenBSD) may choose to make changes to only *their* version and not necessarily report or try to get their changes back into the upstream project. In these cases, the version they review may not be the version all other systems use.

It also notes that experts can be hired to determine whether the developers follow good security practices when developing the software. Signs that good security practices are being followed could include the following:

- The program’s design minimizes privileges (e.g., only small portions of the program have special privileges or the program has special privileges only at certain times).
- The developers strive for simplicity (simpler designs are often more secure).

- The program checks inputs with rigorous whitelists (a “whitelist” defines what is legal input; all other input is rejected).
- Source code scanning tools report few problems when applied to the program.

The section on reliability notes the following metrics:

- Self-reported status (e.g., “mature”)
- Presence of an automated (regression) test suite.

## 7. Doomed to FAIL Index

Tom “spot” Callaway, Fedora Engineering Manager at Red Hat, posted “How to tell if a FLOSS project is doomed to FAIL (or at least, held back...)” in 2009.<sup>4</sup> The handbook *The Open Source Way* includes a chapter with an updated version of this index and is available online [Callaway]. This index is intended to be a quick measure of how well a FLOSS project follows common practices, particularly those that impede packaging or co-development by others. It measures “FAIL” points, so *low* scores are better; 0 is perfect, 5 through 25 is “You’re probably doing okay, but you could be better,” and above 25 is an indicator of serious problems.

The measures are grouped into categories: size, source (version) control, building from source, bundling, libraries, system install, code oddities, communication, releases, history, licensing, and documentation. Examples of causes for fail points are:

- Source Control: There is no publicly available source control (e.g., cvs, svn, bsr, git) [ +10 points of FAIL ].
- Building from source: There is no documentation on how to build from source [ +20 points of FAIL ].
- Communication: Your project does not have a mailing list [ +10 points of FAIL ], or your project does not have a website [ +50 points of FAIL ].
- Licensing: Your code does not have per-file licensing [ +10 points of FAIL ].

Obviously, a high score does not always doom a project to fail, nor does a low score guarantee success. However, like any metric, the score can provide a simple metric to point out potential issues in an OSS project. It is intentionally designed to produce a numerical score, making it relatively easy to report.

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<sup>4</sup> <http://spot.livejournal.com/308370.html>

## 8. Internet Success

The book *Internet Success* by Schweik and English reports a detailed quantitative analysis to determine “what factors lead some OSS commons to success and others to abandonment” [Schweik2012].

Schweik and English examined over 100,000 projects on SourceForge,<sup>5</sup> using data from SourceForge and developer surveys, and using quantitative analysis instead of guesswork. They use a very simple project lifecycle model—projects begin in initiation, and once the project has made its first software release, it switches to growth. Based on this, they classified OSS projects into six “success and abandonment” classes as follows (see their chapter 7 table 7.1):

1. *Success, initiation (SI)*. The developers have produced a first release. Its operational definition is having at least one release (all projects in the growth stage by definition meet the SI criteria, but see below).
2. *Abandonment, initiation (AI)*. The developers have not produced a first release and the project is abandoned. Its operational definition is having zero releases and having  $\geq 1$  year since project registration.
3. *Success, growth (SG)*. The project has achieved 3 meaningful releases of the software, and the software is deemed useful for at least a few users. Its operational definition is  $\geq 3$  releases and  $\geq 6$  months between releases and  $>10$  downloads.
4. *Abandonment, growth (AG)*. The project appears to have been abandoned before producing 3 releases of a useful product, or has produced 3 or more releases in less than 6 months and is abandoned. Its operational definition is 1 or 2 releases and  $\geq 1$  year since the last release, or  $\geq 3$  releases and  $<11$  downloads during a 6+ month time period since the date of first release, or  $\geq 3$  releases in less than 6 months and  $\geq 1$  year since the last release.
5. *Indeterminate initiation (II)*. The project has yet to reveal a first public release but shows significant developer activity. Its operational definition is 0 releases and  $<1$  year since project registration.
6. *Indeterminate growth (IG)*. The project has not yet produced 3 releases but shows development activity, or has produced 3 releases or more in less than 6 months and shows development activity. Its operational definition is 1 or 2 releases and  $<1$  year since the last release, or 3 releases and  $<6$  months between releases and  $<1$  year since the last release.

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<sup>5</sup> <http://sourceforge.net/>

Their operational definition of success initiation (SI) is oversimplified but easy to understand: an SI project has at least one release. Note that their operational definition for a success growth (SG) project is very generous: at least 3 releases, at least 6 months between releases, and has more than 10 downloads.

One of the key results is that during initiation (before first release), the following are the most important issues, in order of importance, for success in an OSS project according to this quantitative data:

1. “Put in the hours. Work hard toward creating your first release.” The details in Chapter 11 tell the story: If the leader put in more than 1.5 hours per week (on average), the project was successful 73% of the time; if the leader did not, the project was abandoned 65% of the time. They are not saying that leaders should put in only 2 hours a week; instead, the point is that the leader must consistently put in time for the project to get to its first release.
2. “Practice leadership by administering your project well, and thinking through and articulating your vision as well as goals for the project. Demonstrate your leadership through hard work....”
3. “Establish a high-quality Web site to showcase and promote your project.”
4. “Create good documentation for your (potential) user and developer community.”
5. “Advertise and market your project, and communicate your plans and goals with the hope of getting help from others.”
6. “Realize that successful projects are found in both GNU General Public License (GPL)-based and [non-GPL] situations.”
7. “Consider, at the project’s outset, creating software that has the potential to be useful to a substantial number of users.” Remarkably, the minimum number of users is surprisingly small; they estimate that successful growth stage projects typically have at least 200 users. In general, the more potential users, the better.

Some items that others have claimed are important, such as keeping complexity low, were not really supported as important. In fact, successful projects tended to have a little more complexity. We suspect both successful and abandoned projects often strive to reduce complexity—so it not really something that distinguishes them—and that sometimes a project that focuses on user needs has to have more complexity than one that does not, simply because user needs can necessitate more complexity.

Additionally, they include guidance for growth projects, which may suggest some metrics. Schweik and English report that, in order of importance, they are:

1. “Your goal should be to create a virtuous circle where others help to improve the software, thereby attracting more users and other developers, which in turn leads to more improvements in the software.... Do this the same way it is done in initiation: spending time, maintain goals and plans, communicate the plans, and maintain a high-quality project web site.” The user community should actively interact with the development team.

We note that possible related metrics include an actively maintained website (e.g., date of last page change on website), messages/month (e.g., email, bug tracker, etc.), number of commits/month, and number of committers.

2. “Advertise and market your project.” In particular, successful growth projects are frequently projects that have added at least one new developer in the growth stage.

We note that possible related metrics include number of developers that have been added (post initial release or within a year).

3. “Have some small tasks available for contributors with limited time.”

We note that a possible metric is a posted list of small tasks for new/limited contributors.

4. “Welcome competition.” The authors were surprised, but noted that “competition seems to favor success.” We do not find this surprising. Competition often encourages others to do better; we have an entire economic system based on that premise.

5. “Consider accepting offers of financing or paid developers (they can greatly increase success rates).” This one, in particular, should surprise no one — if you want to increase success, pay someone to do it.

6. “Keep institutions (rules and project governance) as lean and informal as possible, but do not be afraid to move toward more formalization if it appears necessary.”

They also have tips on how potential OSS users (consumers) can choose an OSS that is more likely to endure. They determined that successful OSS projects have characteristics such as more than 1,000 downloads, users participating in bug tracker and email lists, goals/plans listed, a development team that responds quickly to questions, a good web site, good user documentation, and good developer documentation. A larger development team is a good sign, too.



## C. Specific Potentially Useful Security Metrics

Many more security-focused metrics have been proposed for evaluating software.

### 1. In-Depth Static Analysis Security Tools (e.g., Coverity Scan)

Some tools are specifically designed to look for potential security vulnerabilities and report them. Their sheer counts, perhaps limited to most severe and/or computed as densities, might give an indication of the security (or lack thereof) of software.

Coverity sells a proprietary tool that looks for security vulnerabilities. Coverity Scan, at <https://scan.coverity.com/>, is “a service by which Coverity provides the results of analysis on open source coding projects to open source code developers that have registered their products with Coverity Scan.” It supports C, C++, Java, and C#. An OSS project developer must specifically register their project to participate; results are then sent to the project developers.

The Coverity Scan project was initially launched under a contract with the Department of Homeland Security (DHS) to harden open source software that provides critical infrastructure for the Internet. Coverity Scan began in collaboration with Stanford University on March 6, 2006. During the first year of operation, over 6,000 software defects were fixed across 50 C and C++ projects by open source developers using the analysis results from the Coverity Scan service. DHS support ended in 2009, but the service has continued.

A list of projects covered by Coverity scan is at <https://scan.coverity.com/projects>; over 3,200 participate. Even though the exact results are not posted publicly, the fact that *a project is on the list* maintained by Coverity is public, and that may by itself indicate that a project is interested in detecting and fixing vulnerabilities. A few projects have achieved “rung 2” which is a higher achievement.

A similar argument could apply to other tool makers who make tools that perform in-depth static analysis of software and provide scans of OSS projects. For example, HP/Fortify will provide static analysis tools for examining open source software, in partnership with Sonatype; details are here: <https://www.hpfod.com/open-source-review-project>.

There are some OSS tools that look for vulnerabilities as well. In particular, the *splint* program was designed to do this for C, however, note that *splint* has not been maintained recently. The clang static analyzer does a deeper analysis for buffer overflows and allocation issues in C, C++, and Objective-C programs, but it is not specifically designed for finding vulnerabilities so it lacks many rules for finding them.

Note that these tools use heuristics to determine what a vulnerability is, thus, different tools report different values.

## 2. Lexically Scanning Static Analysis Security Tools (e.g., flawfinder and RATS)

A variant is to use lexically scanning tools to report constructs (“hits”) in software that are of special concern. Again, counts or densities could be reported. OSS tools such as flawfinder and RATS can do this. (Note: David A. Wheeler is the author of flawfinder.)

IDA previously did in-house work measuring hit density, where hits are reports from flawfinder or another lexical tool and density is found by dividing by physical source lines of code. These tools simply report riskier constructs, not really vulnerabilities, but the theory is that if developers often use riskier constructs, they are more likely to produce insecure results. A comparison we did years ago suggests this might be a useful measure. In particular, we found that (at the time) the hit density of the mail transfer agent (MTA) sendmail was significantly larger than that of postfix, and this was consistent with expert opinion of their security at the time.

## 3. Wikipedia Article on OSS Security

Wikipedia’s article “Open-source software security” has various comments about OSS security, including references to metrics and models. The article mentions the following metrics<sup>6</sup>:

- Number of days between vulnerabilities. “It is argued that a system is most vulnerable after a potential vulnerability is discovered, but before a patch is created. By measuring the number of days between the vulnerability [being found] and when the vulnerability is fixed, a basis can be determined on the security of the system. There are a few caveats to such an approach: not every vulnerability is equally bad, and fixing a lot of bugs quickly might not be better than only finding a few and taking a little bit longer to fix them, taking into account the operating system, or the effectiveness of the fix.”
- Morningstar model. “By comparing a large variety of open source and closed source projects a star system could be used to analyze the security of the project similar to how Morningstar, Inc. rates mutual funds. With a large enough data set, statistics could be used to measure the overall effectiveness of one group over the other. An example of such as system is as follows:[7]
  - 1 Star: Many security vulnerabilities
  - 2 Stars: Reliability issues
  - 3 Stars: Follows best security practices

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<sup>6</sup> [http://en.wikipedia.org/w/index.php?title=Open-source\\_software\\_security&oldid=627231105](http://en.wikipedia.org/w/index.php?title=Open-source_software_security&oldid=627231105)  
(permanent link)

- 4 Stars: Documented secure development process
- 5 Stars: Passed independent security review”
- Coverity (see discussion on Coverity).

#### 4. Common Vulnerabilities and Exposures (CVE) Count

MITRE maintains Common Vulnerabilities and Exposures (CVE), a dictionary of publicly known information security vulnerabilities and exposures. MITRE and the National Institute of Standards and Technology (NIST) National Vulnerability Database (NVD)<sup>7</sup> maintain information about publicly known vulnerabilities in released software, including OSS. Not all publicly known vulnerabilities are assigned CVEs, but this is nevertheless a widely used starting point for information about vulnerabilities.

Some obvious metrics suggest themselves:

- Number of CVEs assigned to each particular OSS project, perhaps over some fixed period (say 3 years).
- Number of CVEs with high severity (this is a subset of the whole). The NVD reports Common Vulnerability Scoring System (CVSS) scores for each CVE.
- “Density” version (e.g., by dividing by thousands of lines of code). Since larger projects will tend to have more vulnerabilities, a “density” version can compensate for this.
- Average (or median) number of days between CVE reports.

These have a number of well-understood drawbacks. What we want to know is the number of vulnerabilities *remaining*; however, CVEs report on the number *found*. The current CVE count might be low because no one is looking, or high because substantial effort has been spent to find and report vulnerabilities after it has been released. Also, if projects undergo substantial changes, the CVE counts from older versions may or may not be relevant. Still, if a project has a large number of CVEs, it *might* indicate that the project has not been sufficiently active in countering vulnerabilities.

#### 5. Schryen and Kadura

Guido Schryen and Rouven Kadura in 2009 wrote “Open source vs. closed source software: towards measuring security” [Schryen2009]. In the process they provided summaries of previous work to measure security of OSS in section 3, “Review: Quantitative Models” and provide a metric for measuring software (in their case, to measure responsiveness to vulnerability reports).

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<sup>7</sup> <https://nvd.nist.gov/>

They discuss various models based on security breaches (vulnerability reports). They include various time-based models, although they also note that these become inappropriate when the total effort spent on detecting vulnerabilities is not linear in time. They also review efforts based on software reliability models using exponential equations. Sadly, Rescorla examined empirical analysis of vulnerabilities of both open and closed source operating systems and found no strong statistical evidence that the “G-O” model (a specific form of this approach) approximated the number of detected vulnerabilities over time.

They also note problems in obvious metrics. In section 4, “New Security Metrics,” they note that a single vulnerability that is easy to discover, easy to exploit, and causes severe damage is far worse than 10 vulnerabilities that are extremely hard to discover, can only rarely be exploited, and do not cause significant harm. They note various problems with CVSS, and argue that it would be better to separately categorize and measure a few severity classes (such as high, medium, and low).

Thus, they argue that it is “less reasonable to measure the number of intensity of patches, because this provides no information on the number of covered vulnerabilities or on the ages of covered vulnerabilities. [Instead] compute (statistical data on) the reaction time between detection and elimination of a vulnerability, weighted by the level of severity of the vulnerability. It might also seem reasonable to record how many of the detected vulnerabilities are unpatched.”

They propose a “patch index” metric. This is measured at some time  $t_n$ , producing a metric termed  $PI(t_n)$ , as:

$$\frac{1}{t_n} \sum_{t=1}^{n-1} (t_{i+1} - t_i) \frac{uv_{t_i}}{pv_{t_i} + uv_{t_i}}$$

Where  $i$  is the index of an event that a vulnerability is announced or patched,  $t_i$  is the corresponding point in time,  $pv_{t_i}$  is the (possibly severity-weighted) number of detected and patched vulnerabilities in the time window  $[0; t_i]$ , and  $uv_{t_i}$  is the corresponding (possibly severity-weighted) number of unpatched vulnerabilities. By this measure,  $PI=0$  means that “for all announced vulnerabilities, a patch is already provided at the day of the announcement. In contrast,  $PI=1$  would imply that none of the announced vulnerabilities has been patched.” They show graphs of this metric over time to determine trends.

Fundamentally this metric measures response to vulnerability reports, not the number or severity of vulnerabilities. Their paper shows curves over time of both Microsoft Office and OpenOffice that emphasize this. In their study, Microsoft Office had about seven times more public CVE vulnerability reports than OpenOffice, but the leveled-off patch index is somewhat similar. They note that “probably more vulnerabilities in OpenOffice than in MS Office might have [existed], been detected, potentially discussed in forums, and finally

removed, before they could become a CVE vulnerability.” On average Microsoft Office had 27% of all announced vulnerabilities unpatched compared to OpenOffice.org’s 18%, while vulnerabilities were patched more rapidly in Microsoft Office (median 67.5 days, mean 87 days) than in OpenOffice.org (median 85 days, mean 87.4 days). For more information, see their paper.

Note that [Schryen2011] is also by Guido Schryen.

## 6. “Look at the Numbers” – Wheeler

David A. Wheeler (the author) has, for many years, collected quantitative metrics about OSS in the paper “Why Open Source Software/Free Software (OSS/FS, FLOSS, or FOSS)? Look at the Numbers!” [Wheeler2014n]. The referenced papers are also available (from the same URL) as a Google spreadsheet that lists a number of metrics involving security or quality that people have examined.

The security-related measures that might be relevant for our purposes included:

- [Schrye2011] examines data such as the CVEs in the National Vulnerability Database and uses:
  - Mean time between vulnerability disclosures
  - Medians/standard deviations of the severity values of published vulnerabilities as measured by CVSS
  - Percentage/number of unpatched vulnerabilities after 4 weeks, and their severity. The author found 17.6% (30.4%) of the published open (closed) source software vulnerabilities (in terms of the median) are still unpatched, although this is likely a significant overstatement since if the authors could not find evidence of patching, the vulnerabilities were counted as unpatched.
- Counts of CVEs or counts of critically-important CVEs (e.g., a meta-analysis by Bugtraq)—especially if compared to software with similar functionality.
- Percentage of CVEs that are critically important. For example, Nicholas Petreley’s paper “Security Report: Windows vs Linux” [Petreley2004] focuses on criticality percentages. Criticality is heavily influenced by the effectiveness of various countermeasures, but countermeasures matter.
- Vulnerability response time: Average/median time to respond to vulnerability report and produce a fixed version (e.g., see [Krebs2006a] [Krebs2006b] [Krebs2006c]). Interestingly, one study found that (on average) open source suppliers patch more quickly than closed source ones [Arora2006].

Reliability-related metrics that might be relevant include:

- Failure rates as measured by fuzz testing (note that there are varying approaches to fuzz testing).
- Reliability or average time up under stress testing.
- Total time off-line over a given year (this is hard to apply to a large set of small projects, however).
- Number or density of defects found by static analysis tools. Density of defects is the number of defects divided by the source lines of code (SLOC). Most tools will miss defects (false negatives) and falsely report issues as defects (false positives), but this is still an indicator of relative quality, since if developers are consistently using dangerous approaches, they are more likely to eventually make a security-relevant mistake. If the tool focuses on security vulnerabilities, it should be listed in the security-specific metrics. These include reports by Coverity, Reasoning, and others.
- “Maintainability index” [Samoladas2004], published in *Communications of the ACM*, examined almost 6 million lines of code using this metric. This metric was at one point chosen by the Software Engineering Institute (SEI) as the “most suitable tool for measuring the maintainability of systems with high-quality requirements.” Its purpose is to measure the “size” (by various measures) of individual modules, under the theory that if modules are too big the system is hard to manage. However, this value is very much a heuristic built on other metrics; it is calculated as  $171 - 5.2\ln(\text{avgV}) - 0.23\text{avgV}(g) - 16.2\ln(\text{avgLOC}) + 50\sin(\sqrt{2.4\text{avgPerCM}})$ , where avgV is “average Halstead Volume per module” (a size measure based on the number of distinct operators and operands), avgV(g) is average cyclomatic complexity (a measure of structural complexity), avgLOC is average physical lines of code (excluding blank and comment lines), and avgPerCM is the percentage of lines of comments with respect to the lines of code. See the paper for more. Note that the maintainability index is also mentioned by [Spinellis2009].
- Harvard Business School’s “Exploring the Structure of Complex Software Designs: An Empirical Study of Open Source and Proprietary Code” by Alan MacCormack, John Rusnak, and Carliss Baldwin (Working Paper Number 05-016) reported using the following metrics:
  - Change cost. This measures the percentage of elements affected, on average, when a change is made to one element in the system. A smaller value is better, since as this value gets larger, it’s becomes increasingly likely that a change made will impact a larger number of other components and have unintended consequences (e.g., a value of 17.35% means that if a

given file is changed, on average, 17.35% of other files in the system will be changed, suggesting that they depend (directly or indirectly) on that file).

- “Coordination cost.” This is an estimated cost of communicating information between agents developing each cluster. However, this measure is strongly dependent on the size of the system, so this is not useful for comparing projects of different sizes (and thus is not further considered here).

## **7. Presence of Security Test Suite**

Heartbleed [Wheeler2014h], Apple’s “goto fail” [Wheeler2014g], and many other vulnerabilities could have been detected ahead of time through simple negative testing. That is, identify what should *not* be permitted, and include such tests in the regression test suite (e.g., for every field of a message, include a test with an invalid value (e.g., too high/too low for integers, incorrect length in headers, and so on). Such a test suite should be developed rigorously to cover each field or data type. The presence of such a test suite should increase the confidence that the software resists attack; its absence should raise questions.

## **8. Presence/Absence of Past Security Reviews**

Security reviews are no guarantee of finding all vulnerabilities. Still, if there is no evidence of a past in-depth security review (or *any* security review) that overall increases the risk of security vulnerabilities. Lack of security testing is also an issue.

It is also important to note the limits of any particular test or review. For example, Federal Information Processing Standard (FIPS) 140-2 checks “cryptographic modules,” in particular to ensure that the cryptographic modules’ cryptographic algorithms produce correct outputs given certain inputs. However, FIPS 140-2 explicitly does *not* perform any tests or examinations of cryptographic protocols such as SSL or TLS [Wheeler2014h].

## **D. Specific Potentially Useful General Metrics**

Here are other metrics that are related (e.g., they attempt to predict where defects are especially likely). The theory is that if software has many defects, or other problems, it is more likely to have security problems. These are imperfect predictors, since the only way to know whether something is a defect (or a security defect) is to first compare the software to a specification of what it *should* do. However, if some code is unusually complex, it may be more likely to contain a defect. The evidence that security and quality are interrelated is somewhat sparse. However, it is intuitively sensible, and [Woody2014] provides some evidence for it.

This is a huge field, and not as related to our problem, so this subsection discusses an even smaller part of the field.

### **1. Software Quality Metric Measurement Tools**

Many tools, both OSS and proprietary, can measure various static attributes that may suggest something is more or less likely to contain defects. There are many metrics (such as cyclomatic complexity as described below), and many ways to combine various lower-level metrics into higher-level metrics that might have some meaning.

Examples of proprietary software include that of CAST Software,<sup>8</sup> Semantic Design, and McCabe IQ.<sup>9</sup> OSS tools include CCCC<sup>10</sup> and Eclipse Metrics plugin.

One challenge is that different tools deliver differing results. Lincke et al's "Comparing Software Metrics Tools"<sup>11</sup> compares several tools and shows that different tools often report different values for the "same" measurement. This might seem odd since some of these metrics have definitions published in academic journals, but in practice many of these definitions have ambiguities that result in different values.

### **2. Compiler Warning Flags and Style Checkers**

An alternative approach is to use compiler flags and style checkers to maximally complain about potential issues. Both clang and gcc support many warning flags, for example. These can, again, be divided by KSLOC (Thousands of Source Lines of Code) to give density figures.

### **3. Senior Defect Prediction Measures – McCabe (Cyclomatic complexity) and Halstead**

The McCabe (cyclomatic complexity) and Halstead measures were defined in the 1970s for predicting defects in functions/methods (at the time called "modules") through static analysis of code. They are widely used as static measures, in part because they are well known.

McCabe argued that code with complicated pathways are more error prone. His metrics focus on this. Especially known is Cyclomatic complexity, which measures the number of "linearly independent paths"; many consider a number higher than 10

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<sup>8</sup> <http://www.castsoftware.com/>

<sup>9</sup> [http://www.mccabe.com/iq\\_developers.htm](http://www.mccabe.com/iq_developers.htm)

<sup>10</sup> C and C++ Code Counter, <http://cccc.sourceforge.net/>

<sup>11</sup> <http://arisa.se/files/LLL-08.pdf>



concerning. These *do* measure, in a sense, how much effort is needed to do full coverage in branch testing [McCabe1976].

Halstead argued that code that is hard to read is more likely to be fault prone. He estimated reading complexity by counting concepts, such as the number of unique operators [Halstead1977].

Fenton and Pfleeger have noted a number of problems with static measures, since they are clearly not complete measures [Fenton1997]. Of course, a measure does not need to be complete, just useful.

See [tera-PROMISE] for a discussion of some predictive metrics, particularly the McCabe and Halstead metrics.

#### **4. Test Coverage**

There are various ways to measure the quality of the tests (the “test coverage”) of a regression test suite, and many tools (including gcov/gcc) that can measure them. Especially common measures are statement coverage (the percentage of statements executed by a test suite) and branch coverage (the percentage of branches, both true and false, executed by a test suite).

#### **5. Source Lines of Code**

Many organizations measure source lines of code (SLOC). By itself SLOC says nothing about security, but SLOC is highly correlated to development effort, and it is also correlated to any review effort (although other factors, such as the complexity of those lines, are also important). One complication is that SLOC can be measured different ways: text lines, physical SLOC (which skip blank and comment lines), and logical SLOC (which measure logical statements).

David A. Wheeler’s “sloccount” can automatically measure physical SLOC for many languages. One list of tools is available at <http://www.locmetrics.com/alternatives.html>.

#### **6. Lincke Survey**

Lincke, et al’s “Comparing Software Metrics Tools”<sup>12</sup> looked at many different tools for static metrics for (Java) source code. They identified 17 object-oriented metrics that (1) they could rather securely assign to the same concept, (2) are known and defined in literature, and (3) work on the level of Java classes. They selected nine such metrics that most of the software metrics tools they identified would report. They are:

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<sup>12</sup> <http://arisa.se/files/LLL-08.pdf>

1. CBO (Coupling Between Object classes) – is the number of classes to which a class is coupled.
2. DIT (Depth of Inheritance Tree) – is the maximum inheritance path from the class to the root class [5].
3. LCOM-CK (Lack of Cohesion of Methods) (as originally proposed by Chidamber & Kemerer) – describes the lack of cohesion among the methods of a class.
4. LCOM-HS (Lack of Cohesion of Methods) (as proposed by Henderson-Sellers) – describes the lack of cohesion among the methods of a class.
5. LOC (Lines Of Code) – counts the lines of code of a class.
6. NOC (Number Of Children) – is the number of immediate subclasses subordinated to a class in the class hierarchy.
7. NOM (Number Of Methods) – is the number of methods in a class.
8. RFC (Response For a Class) – is the set of methods that can potentially be executed in response to a message received by an object of the class.
9. WMC (Weighted Methods per Class) (using Cyclomatic Complexity as method weight) – is the sum of weights for the methods of a class.

Of course, a metric might be widely reported because it is easy to measure—not because it is useful. Still, widely measured metrics might be useful.

## **7. Estimating Commit Sizes Efficiently – Hoffmann and Riehle**

A common variable for measuring work contributed is the “commit size,” i.e., the number of lines added, removed, and changed. However, post-facto this can only be estimated; typically, the only information that is available (especially in metrics repositories) is the number of lines added and removed for each change, and obvious ways (such as adding them up) lead to errors.

Philipp Hofmann and Dirk Riehle devised an improved method to estimate commit size based on the data actually available (the number of added and removed lines) using a linear regression model:

```
function real diff_size(int a, int r)
  if ( $0.01269 \times a + 0.01540 \times r > 2.9965$ )
    return  $0.9497 \times a + 0.9744 \times r - 2.9965$ 
  else
    return  $0.9370 \times a + 0.9590 \times r$ 
```

end

end

More information is available in [Hofman2009].

## **8. Choosing Software Metrics – Gao**

“Choosing software metrics for defect prediction: an investigation” [Gao2011] examined how to identify a relatively small set of metrics that nevertheless supported defect prediction (separating modules into “fault-prone” and “not-fault-prone”) [Gao2011]. “The results demonstrate that while some feature ranking techniques performed similarly, the automatic hybrid search algorithm performed the best among the feature subset selection methods. Moreover, performances of the defect prediction models either improved or remained unchanged when over 85% of the software metrics were eliminated.”

The most frequently selected attributes were: “number of distinct include files (FILINCQU), number of different designers making changes (UNQDES), deployment percentage of the module (USAGE), base 2 logarithm of the number of independent paths (LGPATH), total span of branches of conditional arcs (CNDSPNSM), number of problems fixed that were found by designers in the prior release (DESFIX), and number of problems fixed that were found by customers in the prior release (CUSTFIX).”

## **9. Assessing Predictors of Software Defects – Menzies**

“Assessing Predictors of Software Defects”<sup>13</sup> by Tim Menzies et al of 2004 found that “When learning defect detectors from static code measures, NaiveBayes learners are better than entropy-based decision-tree learners. Also, accuracy is not a useful way to assess those detectors. Further, those learners need no more than 200–300 examples to learn adequate detectors, especially when the data has been heavily stratified....” This doesn’t seem as directly relevant for our purposes.

## **10. How Many Software Metrics for Defect Prediction? – Wang**

“How Many Software Metrics Should be Selected for Defect Prediction?” by Wang, et al examine various learning algorithms over various metrics [Wang 2011].

Overall, they found that the best classification model for their dataset (using Eclipse data) was built with only three features selected by the AUC<sup>14</sup> ranker using the logistic regression (LR) learner. Unfortunately, they do not appear to reveal *which* features, nor

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<sup>13</sup> <http://menzies.us/pdf/04psm.pdf>

<sup>14</sup> Area Under the Receiver Operating Characteristic (ROC) Curve

have we found enough data to allow simple reuse of the generated model. This lack of key information makes the work tantalizing but difficult to use or repeat.

## **11. Software Defect Prediction – Punitha**

Punitha and Chitra’s “Software defect prediction using software metrics – A survey” surveyed materials to “help developers identify defects based on existing software metrics using data mining techniques and thereby improve software quality, which ultimately leads to reducing the software development cost in the development and maintenance phase. This research focuses in identifying defective modules and hence the scope of software that needs to be examined for defects can be prioritized.” [Punitha2013] [Mishra2012]

In their approach, they built an inference system to predict modules most likely to be defective using supervised learning (inferring a function from labeled training data). Specifically they applied “SVM,<sup>15</sup> a supervised training algorithm for classification of data into two sets, buggy and non-buggy. Then various rules [are inferred] from the support vectors. The final set of the rules is chosen from the given set of rules using genetic algorithm optimization. The experiments were performed on Eclipse bug data....”

Oddly, the paper describes figures of merit to show their effectiveness, and it seems to suggest that the effectiveness is measured in the paper, but the actual measurement values for software defect prediction do not seem to be in the paper.

In any case, this paper notes the potential use of supervised learning to help determine vulnerable software or modules. Focusing modules may have real advantages—there is more data to work with, and focusing on problematic portions of a program (instead of the entire program) may increase the likelihood of real improvements. However, these are not focused on measuring security, and there is always the risk that a module that doesn’t *seem* like a vulnerable or defective module is implementing a severe vulnerability in a straightforward way.

## **12. COQUALMO**

COQUALMO (COConstructive QUALity Model) is “an estimation model that can be used for predicting [the] number of residual defects/KSLOC (Thousands of Source Lines of Code) or defects/FP (Function Point) in a software product.” Information, and a spreadsheet that implements the model, is available at <http://csse.usc.edu/csse/research/COQUALMO/>.

It was developed in part by Barry Boehm and is similar to the COCOMO effort estimation model. In particular, it requires information on the “defect removal” processes.

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<sup>15</sup> Support Vector Machines

It is not clear that this model would be particularly helpful for our situation, where we must examine a large number of different projects (instead of a single project).

### **13. DoD/Industry 1994 Survey**

David A. Wheeler co-authored a 1994 survey of software metrics in the Department of Defense (DoD) and industry [Springsteen1994]. This is ancient history, but it clearly shows that interest in software metrics has been around a long time. This survey was extremely broad, covering metrics for a variety of purposes not specifically relevant to our case. It made various points, e.g., that “collecting data is different [in general] and it is important to have a simple, goal-directed metrics program” [Springsteen1994, 2.6.1].

Common metrics used in the DoD that are relevant here are:

- Source lines of code (SLOC) for measuring size (with some variation on definition)
- Defect status metrics, e.g., the number and age of unresolved issues.

In industry the most common calculated metric in use was error density. Other metrics relevant to our purposes included:

- Customer severity days (severity of customer problem multiplied by days open, summed by severity level)
- Problems per user-month
- Mean time to defect after release
- Defect containment effectiveness (number of defects removed after internal review but before release, divided by the (number of defects removed after internal review but before release + number of defects remaining in release)).

### **E. Attack Surface Measurement**

A different approach is to try to measure how easy it is to attack a program or system. If software has code that looks like a vulnerability but that cannot be exploited, it doesn't really matter.

“Measuring Relative Attack Surfaces” by Michael Howard, Jon Pincus, and Jeannette M. Wing introduced a metric “for determining whether one version of a system is more secure than another with respect to a fixed set of dimensions. Rather than count bugs at the code level or count vulnerability reports at the system level, we count a system's attack opportunities. We use this count as an indication of the system's “attackability,” likelihood that it will be successfully attacked.” [Howard2003]

“Measuring a System's Attack Surface” by Pratyusa Manadhata and Jeannette M. Wing [Manadhata2004] develops a process for measuring and comparing *systems* (not

individual programs’) attack surfaces. They note that “Today we commonly use two measurements to determine the security of a system: at the code level, we count the number of bugs found (or fixed from one version to the next), and at the system level, we count the number of times a system version is mentioned in...Common Vulnerabilities and Exposures (CVEs) [31] etc. We argue [that] both measurements, while useful, are less than satisfactory....The system actions externally visible to the system’s users together with the system resources accessed or modified by each action constitute the system’s attack surface. Intuitively, the more actions available to a user or the more resources accessible through these actions, the more exposed the attack surface. The more exposed the attack surface, the more likely the system could be successfully attacked, and hence the more insecure it is. We can reduce the attack surface to decrease the likelihood of attack and make a system more secure...certain system resources are more likely to be opportunities, i.e., targets or enablers, of attack than others.... We identify the system resources that are opportunities of attack by a given set of properties associated with the resources, and categorize them into attack classes. These properties reflect the attackability of a type of resource, i.e., some types of resources are more likely to be attacked than other types.” They then provide a specific description of this approach that counts the number of open TCP/UDP sockets, open Remote Procedure Call (RPC) endpoints, services running as root, etc. As written, this is intended for evaluating entire Linux distributions (or similar), not for evaluating individual software packages. It might be possible to identify the packages that directly *cause* these increases in the attack surface, to help identify specific packages involved in the attack surface. This would not identify indirect attacks; dependency information might help but might also pick up too many packages unlikely to be vulnerability sources.

A later paper [Manadhata2007] expands on this approach and focuses more directly on measuring specific applications, e.g., Internet Message Access Protocol (IMAP) servers and File Transfer Protocol (FTP) daemons. That paper works to formalize the notion of a system’s attack surface and proposes “a method to measure a system’s attack surface systematically.” However, it still requires domain knowledge and execution of the program; it is not clear that this would scale well to the large number of programs we are considering in our case.

Open Web Application Security Project (OWASP) has some information on attack surface at: [https://www.owasp.org/index.php/Attack\\_Surface\\_Analysis\\_Cheat\\_Sheet](https://www.owasp.org/index.php/Attack_Surface_Analysis_Cheat_Sheet).

## **F. Kenn White Metrics Set**

Kenn White identified some metrics that might be important (this list was provided to us by the Linux Foundation):

- Project (name)

- Source (URL)
- Maintainers
- Commits
- Notes.

In a separate list he identified the following metrics:

- Project (name)
- Source Repo (URL)
- CoreDependents
- Maintainers
- PrimaryMaintainer
- Affiliations
- CommitsLast30
- StatsInfo (where to get statistics)
- SlocApprox
- IssuesOlder1YR (a count of the number of issues open for more than 1 year)
- OpenIssues (count of open issues)
- IssuesInfo (URL for information on issues).

## **G. IDA Ideas**

### **1. Exposure to Attack**

Some software vulnerabilities are more important than others. A vulnerability that can be exploited externally through a network is especially bad, although a vulnerability that requires execution on a local system (particularly one that allows privilege escalation) is bad as well. Packages directly exposed to network attack are obviously at risk for external exploitation, but indirect attacks are also dangerous (e.g., a vulnerability in an image processing or decompression program might be remotely exploitable).

One remarkably simple approach is to simply note whether a package has at least one CVE. A program that has at least one CVE assigned to it has, by definition, had at least some way to exploit it. Therefore, any program with at least one CVE should be considered as exposed to at least some level.

A simple measure of “external exposure” might be helpful; a program or library that is externally exposed is likely more important to evaluate than one that is not. It is harder to counter attackers who are authorized to run software on a system, so we propose starting

by focusing on countering attack via external networks. Here is a simple proposed qualitative measure (as an example):

- 3: Direct network exposure to attacker. Programs that connect to a network port or directly manage protocols (e.g., OpenSSH).
- 2: Known exposure to external network attacker data because they are often involved in processing of it. This includes decompression algorithms and image processing algorithms. It also includes anything that took user data and transformed it (e.g., turning it into a query) before sending it to a database.
- 1: Known to pass attacker data into the system from external networks, but do not meet the previous criteria. This includes database systems (e.g., MariaDB and the dbm implementations). This also includes any shell used as `/bin/sh`; POSIX and many programming languages include built-in calls that go through the shell. These have a somewhat lower risk than programs that directly process the data, but small errors can still cause serious issues. Note that SQL injection attacks would normally be covered by the above categories, since SQL injections are typically caused by a failure to process attacker-provided data correctly (e.g., using prepared statements), and not by the database systems themselves having a vulnerability. Shellshock is an example of this category; in Shellshock the bash shell *did* process externally provided data in an unexpected way.
- 0: Little external exposure. Attacks might still subvert vulnerabilities in these programs and cause great damage, but that is deemed less likely.

There is also the risk that something could be exploited to cause privilege escalation. This especially includes any package with a `setuid` or `setgid` program (especially if it grants privileges as root or another privileged user), as well as privileged processes that users can communicate with (e.g., via local sockets or Dbus). There is a risk that the programs these depend on could be used; perhaps package dependency information could be used to estimate this.

There is some literature on quantitatively measuring the attack surface of programs. However, the attack surface measurement approaches described in the literature would be time-consuming and difficult to apply across a high number of different programs in different languages, so we have not pursued this approach.

## 2. Other Work

IDA previously did in-house work measuring hit density. Hit density is simply the number of hits divided by KSLOC, where hits are reports of risky constructs (that may lead to vulnerabilities) from `flawfinder` or some other lexical tool. These tools simply report riskier constructs, not vulnerabilities, but the theory is that if developers often use riskier



constructs, they are more likely to produce insecure results. A comparison of sendmail and postfix of years ago suggests this might be a useful measure.

Programming language can also be a potential indicator. A vulnerability can be created in any language. That said, it is especially easy to write vulnerable code in C and C++ because some of their fundamental operations (e.g., array and pointer access) provide no protections (e.g., against buffer overwrites or overreads)—a problem exposed by Heartbleed. C and C++ do not provide automatic memory management; there are advantages to this, but it can increase code complexity and provides additional methods of attack (e.g., through double-frees). PHP<sup>16</sup> is another potentially concerning language, in part because some of its operators have surprising properties (e.g., due to surprising type conversions), but also because many developers with limited skills develop using PHP. Thus, while many good developers use PHP, many developers who do not know how to develop secure software use PHP. David A. Wheeler (one of the authors) recommends specially examining programs written in PHP, simply because so many people who write PHP programs do not know how to do it well.

Programs that contain a large percentage of code developed many years ago (say, more than 10 years ago) can be a potential sign of trouble. Many years ago, fewer developers knew how to develop secure software, and a program with that much unchanged code may suggest that it is fairly inactive. Of course, software can be stable because it is well written and its requirements have not changed, so this measure is not always a sign of trouble. This kind of information can often be retrieved from version control systems, but in some cases this information is not easily available. A plausible proxy could be the project start date; a project that started long ago, even if active, might harbor many vulnerabilities due to old bad practices. For example, some vulnerabilities in X-Windows found in 2014 were due to old code [Thomson2014].

Complexity density (cyclomatic complexity divided by KSLOC) is another potentially promising measure of riskiness. If software is especially complex as measured by complexity density, compared to other software, this could indicate unusually complex software. Since such software is harder to review, it may be more likely to harbor vulnerabilities.

## **H. London January 2015 Meeting**

On January 11 and 12, 2015, the invitation-only “Core Infrastructure Workshop” in London examined what metrics would suggest a project that needs investment. David A. Wheeler participated in this meeting, and many people contributed their ideas of what metrics might be appropriate. Here is a summary created by Wheeler.

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<sup>16</sup> PHP originally stood for Personal Home Page, but it now stands for PHP: Hypertext Preprocessor.

First, metrics that might assess project health:

1. Easy:
  - a. History of many vulnerabilities (e.g., as counted by CVEs).
  - b. Few/1/no developers.
  - c. Poor bug response (long delay on average, large percentage ignored).  
Unfortunately, it is often difficult to distinguish between bug reports and feature requests (there is no standard way to distinguish them, and many projects do not distinguish them at all). When there is no simple way to distinguish between bug reports and feature requests across many projects, merge their results into a single value.
  - d. Number of commits over time.
  - e. Exploitability (e.g., direct or indirect exposure to remote network, potential for exploit use as local privilege gain).
  - f. Language in use (e.g., C or C++).
2. Hard but could be easy:
  - a. Examine dependency information—this can help indicate importance (a program may be very important if many other programs depend on it).
  - b. Differentiate between bugs and feature requests—create a standard way to see the difference across all projects.
  - c. Determine whether it has been fuzz tested or audited.
3. Hard:
  - a. Measure security directly (unfortunately, we don't know how to measure security directly).
  - b. What can projects do to make it easier to capture metrics? (e.g., tweaks in GitHub, SourceForge, etc.)
  - c. Huge impact if broken. This is like importance/widespreadness (e.g., “What percentage of the Internet breaks if this breaks?”).
  - d. Cross-project algorithmic similarity (e.g., if code is copied/pasted, how can we tell that there's a common problem?) The company SourceDNA actively works on this problem.
  - e. Scoring algorithms: How do you combine metrics especially well?

- f. It would be possible to use learning algorithms and data analysis to see what is effective, but that requires that we know the “truth values” (which we typically do not know).
- g. What are the stakeholder priorities?
- h. Exposure (this is related to exploitability).

There was a perceived need to make this data available to the public.

Once higher-risk projects are identified there are many potential responses. These include modifying the software, funding developers and auditors, refactoring it, and rewriting it from scratch.

## **I. Additional Areas for Review**

Time constraints limited the sources we could examine for measurement ideas. Some additional sources we could review further include:

- Examining more sources on “how to release” OSS (i.e., best practices). The failure to apply best practices may indicate problems. The “FAIL” measure is an example of this approach.
- Looking at more search results (e.g., for “Best software defect prediction metrics” and “OSS security evaluation”).
- Reviewing more OSS evaluation methods. Wikipedia has a long list at [http://en.wikipedia.org/wiki/Open-source\\_software\\_assessment\\_methodologies](http://en.wikipedia.org/wiki/Open-source_software_assessment_methodologies).
- Ross Anderson’s paper “Security in Open versus Closed Systems: The Dance of Boltzmann, Coase and Moore,” <http://www.cl.cam.ac.uk/~rja14/Papers/toulouse.pdf>.

## **J. Comments on Metrics**

Some metrics that are especially obvious to collect at first are:

- Project name. This is more complex than you might think. The same name may be used by different projects, and a single project may have more than one name. Project forks complicate this further. In many cases the project URL can be used to distinguish programs because Open Hub and typical Linux packages include this information.
- Source lines of code (SLOC). Larger programs will take longer to audit, so this number is important to help understand the scale of a program.

The following project data seem very promising as metrics:

- Size of development team (committers), e.g., the number of people who committed over the last 12-month period. 0 is especially bad, but 1 is also bad (this is vulnerable to the “hit by the bus” problem). In some cases version control systems don’t report the real facts (if all changes are allocated to the final committer, not to the submitter).
- Activity metrics. Number of commits or change counts, especially over the last 12 months. Ideally this number is stable or increased over the previous 12 months.
- Bug report responsiveness (e.g., average response time for a bug report). The number or percentage of bug reports that have not been resolved after 30 days, the percentage or number of bug reports resolved after 30 days. These should be for true bugs and should not include wish lists/feature requests, although that may not always be possible. There may need to be a separate statistic for percentage of bugs closed as “WONTFIX”; it is easy to close all bugs quickly if you don’t fix any of them. We should probably only care about the statistic for the last 12 months (e.g., of all bugs opened starting this year). We expect that most bug tracking systems are covered by a few repositories or bug-tracking systems, such as GitHub (label “bug”), SourceForge, Savannah, or Bugzilla. If the system does not separate features vs. bugs, we can only report on the combination. On some systems (e.g., GitHub) there is a way to identify bugs, but there is no standard way to use it and/or it is not always used; in that case, report what you can. It would be great if, in the long term, organizations made it easier to separate bugs from non-bugs in a standardized way.

Data suggesting riskiness (likelihood of security problems) may include:

- Dominant programming language. C is considered especially dangerous, with C++ also dangerous.
- Year of project start (or at least, if it is more than 10 years old). If it started long ago, it may have lots of bad practices. Version control databases may not record the entire history of a project, but they can typically report whether it is more than 10 years old. Open Hub does capture “maturity,” which hints at this.

CVE counts are related to security, but that relationship is complicated. After all, if a program undergoes intense audits and has fixed them, it may have many more CVE reports than a critical program that has been ignored. We intend to capture CVE counts, primarily as a way to determine that a program is especially relevant to security; if it has at least two CVEs, it clearly is relevant to security. (One could be a fluke.) See the discussion earlier about CVEs.

Some “project smells” (some inspired by the FAIL metric) suggest that there may be problems:

- Failure to have a (working) website. This is “communication” in the “doomed to FAIL” index.
- Failure to have a mailing list or Internet Relay Chat (IRC) mechanism. This also is “communication” in the “doomed to FAIL” index.
- Failure to have a public source version control repository. If users can download only the final code, there is no opportunity to review the code as it was *before* users were supposed to use it, and there is little opportunity to collaborate. This is “source control” in the “doomed to FAIL” index. The most common OSS version control software today is almost certainly git, but projects could use subversion, mercurial, the long-obsolete CVS, or other software to accomplish this.
- Failure to have a public bug report tracking system. Without this, it is more difficult to determine whether the developers are being responsive to reports, and it is harder for users to determine whether the problem has previously occurred. It is possible to use mailing lists and IRC for bug reporting, but if that is the *only* mechanism, there is a risk that problems (including vulnerabilities) will be forgotten and not fixed. Some projects intentionally delay publicly reporting vulnerabilities until the fix is available, but there are a variety of ways a public bug report tracking system can support this. The public bug tracking system could be implemented using GitHub’s issue tracker, SourceForge’s tracker, Bugzilla, etc.

Finding code repositories is an interesting problem. We can use Open Hub to find human-visible project info, for example, <https://www.openhub.net/p/openssl>. In some cases it points to “Issue tracker” and “Code Locations.” If not, we can look at “browse code” to see whether it refers to a publicly accessible repository (and not just a copy of a final downloadable tarball, e.g., if it ends in “trunk/” it is almost certainly a subversion repository; if it starts with “git:” or ends with “.git” it is a git repository). Debian often records, as part of its copyright information, a URL for a repository, though this may not be current.

It would be possible to examine projects for more complex measurements, especially of the code itself. However, these may take more time to collect, so it might be appropriate to get these measures only once a subset of projects has been identified—if we want to get them at all:

- Percentage of comments (compared to norm for that language). Programs with few comments may be harder to review, and thus vulnerabilities may be easier to miss. However, this metric is not especially informative—a program with many

comments may be strewn with vulnerabilities, and comments can often mislead. This data is automatically provided by Open Hub, so we may as well use it.

- Potential security defect density. Run Coverity scan, flawfinder, RATS, and/or some other tool and divide these by the KSLOC. Higher numbers suggest that the program often uses risky constructs, and thus is at higher risk for vulnerabilities.
- Warning density. Run warning flag/quality tools (e.g., clang warning flags) and divide by SLOC, again, to see whether the program often uses risky constructs.
- Regression test coverage (e.g., statement coverage, or even better branch coverage). A program that has poor test coverage has many untested areas. This requires running the regression test suite; that is not an easy thing to do for a large set of programs.
- Complexity (McCabe). Complexity measures for functions/methods can be rolled up in various ways (e.g., the percentage of functions or methods with a complexity more than 10, or the average complexity density (for each function, divide by lines of code, report the average)). Highly complex code is more likely to have vulnerabilities. Some OSS tools can collect this data for some languages.
- Vulnerability repair speed, (i.e., responsiveness (how quickly they respond to a vulnerability report, especially a public one)). This may be determinable from the bug database, but this is often harder because bug databases don't always link to vulnerability reports. The CVE database may help.

### 3. Important OSS Projects

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Another challenge is to identify OSS projects that might need investment. These potential projects need to have metrics collected about them, and then those metrics compared to determine which projects to evaluate.

Identifying important OSS projects involves many challenges. Many OSS projects themselves depend on other projects, which may be copied into them; this means that some lower-level libraries are not necessarily obvious as being widely used.

Some OSS projects are rarely used; others are very widely used. In general, we prefer ones that are widely used in various configurations (e.g., OSS that are widely used in common Linux distributions) especially if they are in the minimal install or in common configurations (e.g., server installs, are of special interest).

Not all OSS projects are equally exposed to attack; some software, such as network programs and decompression libraries are obviously directly exposed to data from attackers, while others are less exposed (and thus it is harder to determine their importance for investment). Some lists (e.g., the Linux Foundation original list) take this into account; in addition, we have contemplated using a rough “exposure” metric to capture this issue.

Some sources that can be used as a starting point for identifying these projects are discussed below.

#### A. Common in Distributions

One approach is to take a common Linux distribution (e.g., Debian) and find:

1. Packages always installed in the minimal (“Base”) install
2. Packages installed by a predefined group (e.g., “Web Server”)
3. Packages commonly installed in common cases as a server OS
4. Packages commonly installed by a developer. For example, GnuPG is an infrastructure component widely used to secure the development and distribution process, because it is used for signing in git and many other tools.

We started with the first two (Debian “Base” and “Web Server”), and augmented that with an installation of ssh, because these are extremely common programs. Debian is a useful place to start, since it has a large number of packages that represent OSS widely.

It would be possible to cross-reference this among Red Hat-derived distributions. They sometimes choose different package names, but in those cases, the project URLs could be used to match the actual programs. In many cases Debian and Red Hat-derived distributions use the same underlying programs.

## **B. Linux Foundation Original List**

The Linux Foundation CII frequently asked questions (FAQ) listed the following as “critical open source software projects”<sup>17</sup>:

- Network Time Protocol (NTP); note that there are several implementations of the NTP protocol
- OpenSSH
- OpenSSL.

The Linux Foundation had a brainstorming session to identify some projects that might be worth considering, which resulted in the following list:

1. Compression libraries, including LZ0, LZ4, libgz. We would add implementations of unzip; many formats are basically extensions of the zip format, so there are many routines that depend on them.
2. Pluggable Authentication Modules (pam). Everyone uses it, and it is critically important.
3. Web services libraries: libcurl, libxml, json-c, libpng.
4. Voice: libzrtp.
5. Apache-related: mod\_ssl, mod\_tls, mod\_auth\_\*, mod\_compress.
6. Encryption libraries: LibreSSL, modssl, modtls, BouncyCastle, gpg, otr, axolotl.
7. Static analyzers: Clang, Frama-C.
8. Nginx.
9. OpenVPN. It was noted that the funding model may be similar to OpenSSL, where consulting funds the company. It was also noted that OpenVPN needs to correctly use OpenSSL in order to be secure, so focusing on OpenSSL may be more worthwhile.
10. OpenWRT. Imagination is working on OpenWRT, but may not be focused on security. It was noted this that this would be a big effort.

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<sup>17</sup> <http://www.linuxfoundation.org/programs/core-infrastructure-initiative/faq#faq11>



11. BIOS projects: coreboot and TianoCore.
12. Remote Direct Memory Access (RDMA) – these are kernel drivers that access memory directly (used to support, e.g., InfiniBand).

### **C. Kenn White List**

Kenn White developed a list of potential projects, including:

- libBFD (binutils; gdb)
- libcURL
- libIDN
- libXML
- libLZMA.

Example dependents for Web Servers/Core WS include:

- Apache (2.x): mod\_php, mod\_python, mod\_ssl
- BusyBox
- Django (PycURL)
- FastCGI/lighttpd
- Grails
- Java (JSP, J2EE, XML services)
- Nginx
- PHP (5.x)
- Ruby Gems
- Ruby on Rails (Passenger)
- Tomcat (6/7.x).

Example dependents for Core OS Services are:

- b43 (wireless networking)
- cryptsetup – luks (volume/disk encryption)
- device – mapper
- dhclient (DHCP)
- dracut (bootup, kernel bootstrap)
- Internationalized characters/Unicode/Puny/DNS

- Iproute
- Iptables
- gnupg2
- lvm2
- openssh (server and clients)
- SE Linux (policycoreutils)
- Mail (postfix)
- RPM and Yum
- RSyslog.

#### **D. Google Application Security Patch Reward Program**

On October 9, 2013, Google announced a program to reward proactive security improvements to some open-source projects.<sup>18</sup> As of December 8, 2014, this program limits potential awards to the following OSS projects, which Google appears to consider important:

1. Open-source foundations of Chrome and Android: Chromium, Blink, Omaha, Android Open Source Project (AOSP)
2. Security-critical, commonly used components of the Linux kernel, including Kernel-based Virtual Machine (KVM)
3. High-profile web and mail servers: Apache httpd, lighttpd, nginx, Sendmail, Postfix, Exim, Dovecot
4. Other high-impact network services: OpenSSH, OpenVPN, BIND, ISC DHCP, University of Delaware NTPD
5. Core infrastructure data parsers: libjpeg, libjpeg-turbo, libpng, giflib, zlib, libxml2
6. Other essential libraries: OpenSSL, Mozilla NSS
7. The reference implementation of Certificate Transparency and its open-source dependencies
8. Toolchain security improvements for GCC, binutils, and llvm
9. Security-relevant bits of common package managers: yum, apt, pip, npm

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<sup>18</sup> Details are available at: <https://www.google.com/about/appsecurity/patch-rewards/>

10. Popular web frameworks: Angular, Closure, Dart, Django, Dojo Foundation, Ember, GWT, Go, Jinja (Werkzeug, Flask), jQuery, Knockout, Struts, Web2py, Wicket.

## E. Recent Problem Reports

Some software with known recent problems (vulnerabilities or known lack of support) include:

- Network Time Protocol (NTP) services, since these are directly exposed to external exploitation<sup>19</sup>
- GNU Privacy Guard (GPG), which has only one developer but is widely used to secure software (including signing of commits).<sup>20</sup> There are also known problems with GPG: they have dropped the ability to read old formats (requiring people to keep old versions with known problems) and do not support smart cards.<sup>21</sup>

## F. Augmented List of Programs

As noted earlier, we started with Debian “Base” and “Web Server” augmented with an installation of ssh.

We then looked for programs/projects that should be added. We did not want to include simple bindings for different programming languages; there are many bindings for a given library, and since the functionality is primarily in the base, we wanted to start by focusing on the base library. For our purposes, we use Debian package names as an index key. The categories (based on the above) that we added are:

1. Encryption libraries/tools: These were found using “grep -E '([Ee]ncrypt|[Dd]ecrypt|cryptographic|<TLS>|<SSL>)' ...” on the full Debian package list and then manually reviewing the results. Note that LibreSSL also would qualify. We identified 214 packages, many of which are rarely used or are implementations for specific languages, which makes it harder to determine what to include. Manual review of this list identified the following as potentially especially important: coolkey, gnutls-bin (and other gnutls), libopencryptoki0, and libpolarssl-runtime. Note that libssl1.0.0, openssl, and libgnutls26 were already on the list. Others that may also be important are libace-inet-ssl-6.0.3, aespip, aolserver4-nsopengssl, libbeecrypt7, ccrypt, claws-mail-smime-plugin, courier-imap-ssl, and courier-ssl.

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<sup>19</sup> <https://ics-cert.us-cert.gov/advisories/ICSA-14-353-01>

<sup>20</sup> <https://gnupg.org/blog/20141214-gnupg-and-g10.html>.

<sup>21</sup> <https://lwn.net/Articles/626660/>

2. Compression libraries. The Debian packages in the base that have “compress” in their description are: bzip2, gzip, libbz2-1.0, liblzma5, xz-utils, and zlib1g. (These are thus already covered in our starting set.) At a minimum, adding “zip” and “unzip” (from info-zip) seems wise.
3. Image processing libraries.
4. C/C++-based XML processing libraries.
5. Key network protocols: ntp (Network Time Protocol), bootp.

We then added others as they were identified as important so they could be analyzed. Our time was limited; adding more programs would be useful.

The story of PAM<sup>22</sup> is complex. The OpenPAM program is used by many \*BSDs<sup>23</sup> and Apple MacOS X; see <http://www.openpam.org/wiki/History> for details. This is not the same as Linux-PAM, which is used by many Linux distributions. Thus, multiple projects may appear to have the same name, which can be confusing. In addition, the site at <https://linux-pam.org> cross-links to <https://fedorahosted.org/linux-pam/>, and on both sites at the time we examined it several links did not work that should have led to more information about the development status. Our data about Linux-PAM suggested it was not very active, but this is probably an artifact of the difficulty that the automated tools had in tracing the web sites to determine the actual situation.

The library for Binary File Descriptor (BFD) files is part of binutils; binutils itself has significant maintenance, but the BFD portion less so. This is difficult to tease apart because of the way it is packaged.

Kenn White mentioned “otr”; we believe this is the “off-the-record” (otr) message protocol, and probably more specifically the portable sample library and toolkit implementation available via the home page <https://otr.cypherpunks.ca/>.

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<sup>22</sup> Pluggable Authentication Module.

<sup>23</sup> \*BSD is a conventional abbreviation for the operating systems that descended from the Berkeley Software Distribution (BSD), including FreeBSD, NetBSD, and OpenBSD.

## 4. Selected Approach and Early Results

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### A. Overall Approach

We had little time to complete the analysis, and the goal was to quickly identify plausible candidates for human review. Thus, we focused on metric values that can be easily and quickly acquired. In particular, we used data from Black Duck Open Hub where possible, as well as from the Debian package repository, because these are easy to get quickly.

We plan to extend this start with other information that can help us quickly filter out likely candidates. We are gathering this data via programs so that the results can be re-run later as more data is desired or updated information becomes available.

We have been asked by the Linux Foundation to focus on projects that have relatively little activity or current development effort. For example, the content management systems (CMS) WordPress, Joomla!, and Drupal are widely used, and all have had many vulnerabilities identified (especially when their plug-ins are included).<sup>24</sup> However, these CMSs have a number of developers behind them, who already look for and attempt to counter vulnerabilities. They could do better, but at least there is typically progress in those areas; our concern for now is those projects that are relatively inactive and thus are unlikely to improve over time. Some CMS plug-ins are widely used yet are inadequately audited; we are not looking at such plug-ins at this time, but they would be good candidates for future investigation.

### B. Caveats

We have developed software to automatically gather data, attempt to “clean it up” (e.g., to match names), and report it. However, this is subject to the messiness of real-world data.

Names in particular are difficult. The same name may be used by unrelated projects, forks of projects may have different names, and different sources may use different names for the same project. We have tried to correlate the data, but this is certain to be imperfect.

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<sup>24</sup> Those who doubt this can just look at <http://cve.mitre.org/cgi-bin/cvekey.cgi?keyword=wordpress>, <http://cve.mitre.org/cgi-bin/cvekey.cgi?keyword=joomla>, and <http://cve.mitre.org/cgi-bin/cvekey.cgi?keyword=drupal>.

Open Hub is an important data source for our process, but in some cases it provides only summary data in its API (e.g., it reports that the number of commits is going up or down, but not the exact number). In many cases we can get this information from other sources, but we did so only when it seemed important.

The first CVE reports used keyword matching. This means, for example, that if a CVE report mentions product X, then the CVE will show up even if the CVE does not actually apply to product X. CVEs are often specific to environments. This was later changed to use the Debian project's special mapping database, which greatly improved the results.

Since it is difficult to directly estimate the vulnerability of software, we primarily emphasized process measures that suggest inadequate maintenance, as described above.

Our primary purpose is to identify likely OSS candidates for investment. It is acceptable to have some false positives (that is, to identify some candidates that do not really need significant investment), because the list of candidates will then be reviewed by humans. Our primary goal was to avoid significant false negatives, that is, we wanted to avoid failing to identify a candidate that truly needed investment. As a result, we tended to identify a candidate as needing investment when we were not sure (e.g., due to lack of data about the candidate). However, we could not take this to extremes because identifying all OSS projects would not be useful.

## **C. First Stage**

On December 16, 2014, we submitted an early mock-up of OSS projects and data about them. This was partly created automatically, and partly by hand, to see whether our overall approach would work. We determined that although there were difficulties with data “messiness,” the overall approach did work.

We first took the list of Debian packages in Base + Web-server + ssh. Specifically we used Debian 7.7.0 (Wheezy) stable version for the x86\_64 platform; this was first released on October 18, 2014. That has 368 binary packages (not including 3 “tasks” that are not really packages, and including 4 packages that are only installed when running on virtualbox). Our theory is that if a package is in the Debian base, it applies to most Linux distributions (it almost certainly applies to Ubuntu, and Red-Hat-based systems have many of the same software packages). The packages probably apply to many non-Linux situations as well. In some cases, one source package generates multiple binary packages. We can expand this, of course, but we think it's a plausible start.

We then cross-referenced that list to Open Hub metrics. There is about a 70% match (about 70% of the time, we found that Open Hub has metrics data), which saves a lot of time. We had to hand-determine the mapping between Debian and OpenHub. In the longer term, matching on home page URLs might help.

OpenHub does not provide all the data we would like to have (e.g., the exposure to vulnerabilities or how well they process bug reports) but it still helps us identify projects that are probably not well maintained.

We then investigated whether just those metrics would be adequate for finding concerning projects. We filtered on projects that Open Hub reported were Small/Single Developer and Small/Stable Activity. Note that this automatically does not include the 30% of projects for which we have no data, but we were basically seeing whether the process made some sense. We then manually looked through the list. Just doing that suggested plausible candidates, such as (giving the Debian package names): PAM-related packages (libpam-modules, libpam-modules-bin, libpam-runtime, libpam0g), libexpat1, procmail, zlib1g, libsasl2-2, libsasl2-modules, gzip, libfuse2, libgpgme11, libkeyutils1, and libpng12-0. Ones that met the criteria, and perhaps might justify more investigation, included: netcat-traditional, dmidecode, libgdbm3, libnfnlink0, libsemanage1, libsemanage-common, locales, and libpopt0.

As noted, there are projects that we don't have any data on, and there are other important metrics that OpenHub doesn't provide (bug tracker or repository URL, CVE data, bug processing data, whether or not there are setuid or accessible privileged programs, etc.). In addition, there were almost certainly projects that were not in that set of Debian packages that should be considered. However, this first pass convinced us that we were making reasonable progress and that our approach was reasonable.

The heading names began with “Debian\_” if the data source was Debian, and “Openhub” if the data source was Open Hub (Ohloh).

The headings in this version were:

1. Debian\_Package: Debian name of binary package
2. Debian\_Source: Debian name of source package; one source package may generate multiple binary packages
3. Debian\_Version
4. Debian\_Description: Debian's one-line description
5. Debian\_Homepage: Homepage for the entire project
6. Debian\_Install: Debian installation package, e.g., Standard System Utilities
7. OpenHub Query Name: This mapped the Debian package name to the name useful for querying; in some cases this is odd, e.g., Debian's “apt-listchanges” maps to “8066”
8. Openhub\_name: The user-visible name produced by Open Hub
9. Openhub\_description

10. Openhub\_homepage\_url
11. Openhub\_download\_url: Location where the stable version (e.g., tarball) can be downloaded
12. Openhub\_twelve\_month\_contributor\_count: Number of contributors within the last 12 months
13. Openhub\_total\_contributor\_count: Total number of contributors according to Open Hub; note that many old projects will under-report, since this information is captured from version control systems that may not have full histories
14. Openhub\_total\_code\_lines: SLOC
15. Openhub\_main\_language\_name: Primary programming language
16. Openhub\_licences
17. Openhub\_codebase\_factoid: Length of time extant (as text) (e.g., “Mature, well-established codebase” or “Well-established codebase” or “Short source control history”)
18. Openhub\_activity\_factoid: Activity as measured by commits compared to previous year, e.g., “Increasing Y-O-Y development activity” or “Decreasing Y-O-Y development activity” or “No recent development activity”
19. Openhub\_comments\_factoid: Percentage of comments in source (e.g., “Very few source code comments” or “Few source code comments”)
20. Openhub\_devteam\_factoid: Size of development team (e.g., “No recent development activity” or “Only a single active developer” or “Small development team”).

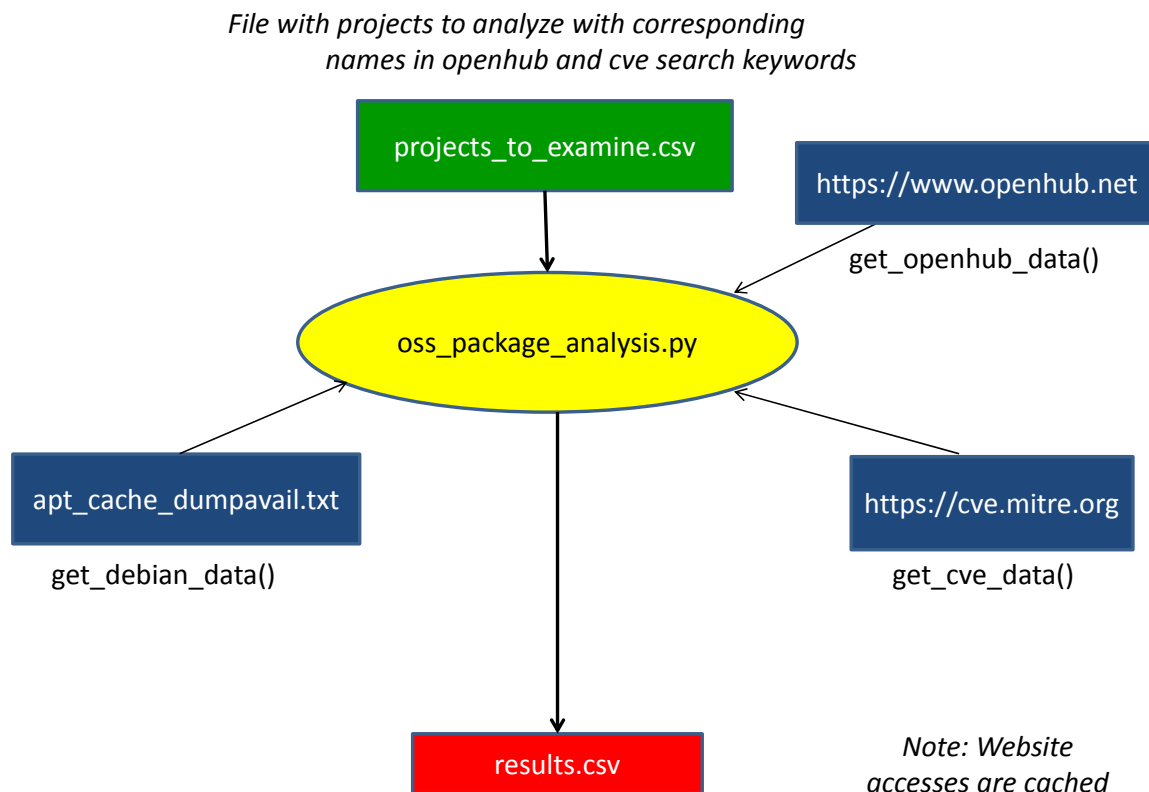
## **D. Second Stage**

We took our early code and modified it to more automatically generate the results (the first version was partly done by hand to see whether the overall approach made sense). Our automated code is implemented in Python, which extracts data from a variety of sources. Data is cached to reduce the impact on servers and to speed updates of results.

We also spent time finding additional matches between the Debian and Open Hub lists; in many more cases we found more matches. We then worked to extract CVE data for each project because this helped us identify what was relevant for security.

The second stage was used to create the draft results shown in London in January 2015.





**Figure 2. Analysis Process, Stage 2**

The “projects\_to\_examine.csv” file lists the projects to examine, along with related information (e.g., the keys to use with CVE or OpenHub). The list of projects to consider began with all files in Debian base, and we then added others as appropriate. This file is editable; we added comments as appropriate to explain why the additional files were added.

We also created an early scoring mechanism to try to combine the metrics. This was a “risk” score, where a higher value suggests a higher risk. This draft combined score is:

- Project website: 1 point if there is no identified project website (identified by either the Debian database or OpenHub database); there may be a website not identified by our databases, but this can be determined later
- CVE vulnerability reports: 3 points if 4+, 2 points for 2-3, 1 point for 1
- main\_language\_name: 2 points for C or C++; secure programs can be written in these languages, but it is especially easy to make vulnerabilities in them
- twelve\_month\_contributor\_count: 2 points for 0 contributors, 1 point for 1-3 contributors
- FactoidTeamSize: 2 points for “No recent development activity” or “Only a single active developer”; 1 point for “Small development team” or unknown.

It's arguable that the last two items double-count the number of contributors.

This process identified several projects that we agreed were concerning (e.g., xauth). However, for some of the projects identified, this scoring mechanism did not do a good job identifying projects with a significant security impact. We believe that part of the problem is that the *exploitability* of a program was not captured at all. Thus, we determined that adding data about exploitability would help us identify relevant programs.

On January 6, 2015, we received approval from Black Duck to use the Open Hub data in a different way than was usually allowed. They simply wanted to be sure that we would not download and redistribute their database wholesale, and they also wanted to ensure that they received credit (which we are happy to provide). They requested the following attribution: "Data sourced from the Black Duck Open HUB, a free online community resource for discovering, evaluating, tracking and comparing open source code and projects." In the event that the data is quoted in text, they asked us to identify the source of that data by referring to it as "Black Duck data" or "data from Black Duck."

At the January 2015 London meeting we received helpful feedback. Many noted the need to stress exploitability or importance (this was consistent with our thinking). Florian Weimer identified a much better source for package-specific CVE information, as well as a popularity database that might be useful. At the same time, Samir Khakimov determined that the Python program would be easier to understand if it used a more OO-based approach (our analysis program had previously grown as necessary to gather data, and we had not focused on the temporary data structures used to capture the data). David A. Wheeler noted that the code should pull all data into memory, or at least into quickly searched data structures, so that results could be quickly recalculated. Quick recalculation made it easier for us to refine our approach.

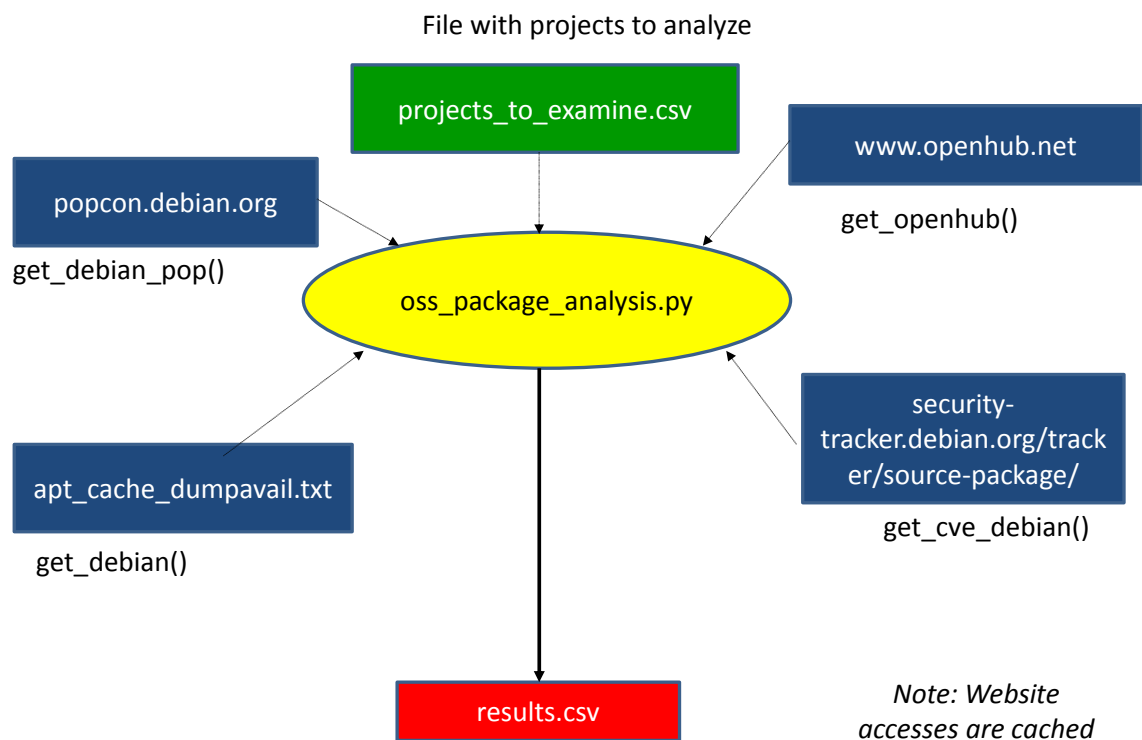
## 5. Current Process

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The latest version of our analysis program is written in Python. It looks at the file “projects\_to\_examine.csv” to determine what to examine, gathers data from a variety of sources, and merges it into a file called “results.csv” for review. Much of this code was written by Samir Khakimov at IDA. The code is released under the MIT license, which is an OSS license. The package names are the Debian binary package names; the list includes the base Debian packages and others that have been identified as important. This is a small set of the over 37,500 packages in Debian.

### A. Current Simplified Approach

Figure 3 shows how the program currently works.



**Figure 3. Analysis Process, Current Stage**

Note that some of this data (OpenHub) is sourced from the Black Duck Open HUB, “a free online community resource for discovering, evaluating, tracking and comparing open source code and projects”; we gratefully acknowledge them. Other sources include

the Debian package information (obtained from Debian’s apt program), popularity data maintained by Debian, and CVE data provided in yet another Debian location. This version responds to January 2015 feedback in London (e.g., adding popularity score and an improved data source for CVE counts).

Unfortunately many of our data sources are extremely noisy. It is often difficult to determine whether there is a match between the Debian database and the OpenHub database (when projects refer to the same home page, there is a clear match, but in other cases it is more difficult). In some cases there is no OpenHub entry at all. Our software works to counter this noisiness, and we hand-reviewed results with higher risk indexes (to reduce the likelihood that the report included irrelevant work).

## **B. Risk Index**

Our revised risk index is:

- Project website: 1 point if there is no identified project website (identified by either the Debian database or OpenHub database). This is given only 1 point because our data sources often fail to identify websites even when they exist.
- `main_language_name`: 2 points for C or C++. Secure programs can be written in these languages, but it is especially easy to make vulnerabilities in them. OpenHub data is used to identify the primary language in many cases; where it is not available, the Debian repository website is scraped to determine the language.
- CVE vulnerability reports: 3 points if 4+ , 2 points for 2–3, 1 point for 1. The CVE count is now direct from Debian, and thus more reliable. CVE counts are a double-edged sword. The number of reports may be low because there are few existing problems or because few reviewed it; the number may be high because there are many existing problems or because the software has undergone extensive review. We are using CVEs primarily to help determine the exposure of the program to attack; if several CVEs exist, then it is clearly exposed to attack.
- `twelve_month_contributor_count`: 5 points for 0 contributors, 4 points for 1–3 contributors, 2 points if the number is unknown (blank).
- Debian popularity count: 1 point if the “popularity” score per Debian is more than the tenth percentile of the packages being analyzed. The set of packages being considered is heavily skewed to packages in wide use, and the “popularity” score is noisy anyway (e.g., a single install from Debian by a router supplier might result in millions of uses). Thus, this is primarily used to reduce the rating of projects that appear to be less used.
- Exposure values: 2 points if directly exposed to the network (as a server or client), 1 point if it is often used to process data provided by a network, and 1 point if it

could be used for local privilege escalation. These values were from expert estimation by the authors.

- Application data only: **Subtract** 3 points if the Debian database reports that it is “Application Data” or “Standalone Data” because this indicates the package isn’t really code but is instead application data for code (e.g., “geoip-database”). We believe there is an error in the Debian dataset; “apache2.2-bin” (“binary”) is marked as application data, while “apache2.2-doc” is not. This error doesn’t matter since binary package “apache2-utils” also maps to the source package “apache2,” and so Apache is still considered. Note that “isc-dhcp-common” is marked as data, but “isc-dhcp-client” is not, so the source package isc-dhcp is still considered.

The risk rating “ca-certificates” (a list of certificate authorities) is lowered by this rule. Since the list of certificates is important for security, this is a debatable result of this rule. However, all major distributions have an active separate process for reviewing their list of certificate authorities. Thus, we do not think this lowering of the ca-certificates index rating is a serious problem.

Areas to potentially improve include:

- Improve/fill in missing data. In general we have messy and incomplete data sources, which require time to address. We already do this in some cases (e.g., to determine the primary implementation language).
- Gather and analyze bug report processing (e.g., how long (on average) does it take to respond to a bug report, and how many bug reports lie unresolved after some time (such as 90 days)). This turns out to be hard data to gather across a large number of projects, because many projects do not separate bug reports from enhancement requests. The “isitmaintained.com” site can analyze GitHub projects to separate bug reports from enhancement requests, but it cannot analyze projects on sites other than GitHub, and it requires that a project use one of the tags it knows about.
- Perform static analysis on source code to determine the likely number of latent vulnerabilities (e.g., using Coverity scan, RATS, or flawfinder); measures such as hit density could indicate more problematic software.

There are many ways to acquire more data, and many ways to combine data to find a combined score. The heuristic above described here is our current attempt.



## 6. Current Process Results

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The file “results.csv” contains the entire set of packages we’ve measured, sorted by the heuristic risk index we created. We first present the list of packages that are considered riskiest, considering only the risk score. We manually reviewed the packages that scored riskiest, and selected the ones we believe *actually* are riskier (since humans can estimate other factors such as the likelihood that a program defect will lead to a vulnerability).

### A. Riskiest OSS Programs (straight from scores)

The OSS Debian binary packages with the largest two risk values are shown in Table 3.

**Table 3. Riskiest OSS Programs (straight from scores)**

BINARY PACKAGE NAME	SOURCE PACKAGE NAME (IF DIFFERENT)
ftp	netkit-ftp
netcat-traditional	netcat
tcpd	tcp-wrappers
whois	
at	
libwrap0	tcp-wrappers
traceroute	
xauth	
bzip2	
hostname	
libacl1	acl
libaudit0	audit
libbz2-1.0	bzip2
libept1.4.12	libept
libreadline6	readline6
libtasn1-3	
linux-base	
telnet	netkit-telnet

## B. Riskiest OSS Programs

We examined the top packages as determined by the score and selected a subset we thought were especially risky. In this process, we used the score to guide us on what to look at, but we also took into account our knowledge of how the programs are used and marked those that appeared most risky with “++”; where we were especially concerned, we marked them with “+++”. Here the “Debian source” is the name of the source package (which is the name of the overall project this comes from), while “project name” is the Debian binary package (a source package can generate many binary packages). We did analysis on a binary package level since there can be distinctions between different binary packages, but the table below re-groups the information into source package names to make this easier to follow.

Projects that we find *especially* concerning, because they appear to be relatively unmaintained even after searching for more information, are the following (we marked these with “+++”):

- bzip2
- gzip
- expat (libexpat1)
- zlib (zlib1g)
- libjpeg8
- libpng (libpng12-0)
- unzip
- mod-gnutls (libapache2-mod-gnutls).

The “libapache2-mod-gnutls” package is especially concerning; this is a key glue mechanism between Apache and GnuTLS, but it seems to be minimally maintained. The others are libraries for processing various formats (images, compression, and XML) that are the basis for many other functions that do not seem to be well maintained. These packages appear because our metrics focus on projects that have relatively little maintenance and either are exposed to the network or are often used to directly process data provided by potential attackers.

Table 4 lists the riskiest OSS programs, as determined by humans examining the list of the projects ranking higher on our risk index. The “comment\_on\_priority” field is ASCII text commenting on why they are prioritized; in some cases words are emphasized through capitalization, and short incomplete phrases are used for brevity. These are sorted by our risk index:



**Table 4. Riskiest OSS Programs (human-identified subset informed by risk measures)**

debian_source	project_name	comment_on_priority
xauth	xauth	X Authentication - might enable local privilege escalation. Home Page: <a href="http://www.x.org/">http://www.x.org/</a> . There are current efforts to replace this (e.g., Wayland, Mir). One nice thing... <a href="http://www.x.org/wiki/Development/Security/">www.x.org/wiki/Development/Security/</a> . ++
bzip2	bzip2	Bzip2 is widely used for decompression of network-supplied data. No code repo, only released code source tarballs. This is the reference implementation, many alternate implementations shown on <a href="https://en.wikipedia.org/wiki/Bzip2">https://en.wikipedia.org/wiki/Bzip2</a> . Last release in 2010, v1.0.6, in response to CVE-2010-0405. No forums found. At least one HTTP server supports bzip2 for compression. +++
audit	libaudit0	Audit framework for Debian, many other components depend on it. Useful for detecting problems. Project infrastructure at Red Hat. SVN repo available. Mailing lists and IRC exist. v2.4 release 10/2014, plans exist for v2.5 and 2.6. Couldn't find an issue tracker. ++
bzip2	libbz2-1.0	Bzip2 is widely used for decompression of network-supplied data. ++
libtasn1-3	libtasn1-3	Important because X.509 certificates (standard for certificates) use ASN.1. A Gnu project, git repo and mailing list available from home page. No issue tracker found. Search on 'security' in mailing list archives shows a few messages from Debian using security-related compiler flags. Debian security page shows a history of CVE's being fixed, most recent in 2014. OpenHub page at <a href="https://www.openhub.net/p/libtasn1">https://www.openhub.net/p/libtasn1</a> gives stats. Active effort of one developer, with some external contributions. ++
bind9	bind9-host, dnsutils, libbind9-80, libdns88, libisc84, libisccc80, libiscfg82, liblwres80	Bind9 is critical for security. Support is via mailing lists. Very active development, issue tracker intentionally not public...claimed to be "huge backlog" of issues. Public git repo. Active history of CVEs raised and fixed. ++
exim4	exim4, exim4-base, exim4-config, exim4-daemon-light	This is an MTA, thus has to process untrustworthy data. Mailing lists, bugzilla tracker, public git repo. Active history of CVE issues and fixes. Active with 16 contributors in last year. ++
isc-dhcp	isc-dhcp-client	In wide use, accepts data from external sources. Includes dhclient. Very similar to BIND. Perhaps slightly less active, lower developer count, but steady commits and releases. Somewhat critical due to being core network functionality. ++
gnutls26	libgnutls26	TLS implementation. Less used than OpenSSL, but is in use (it avoids OpenSSL licensing issues). Good "security activity" metrics, except suspiciously empty GitLab issue tracker, especially when Debian Security page shows many

debian_source	project_name	comment_on_priority
		existing vulnerabilities. They may sweep the public issue tracker so as not to expose what they're trying to fix before it's ready. Notably still supporting SSL 3.0, which has known exploits. There is evidence they do stay on top of things, though. They have responded to questions regarding FREAK in their mailing lists, explaining gnutls is not vulnerable due to how it operates. ++
gpgme1.0	libgpgme11	Vital for protecting email (see gpg) Git repo available. Issue tracker site uses self-signed cert; it's concerning that a user would have to compromise the HTTPS security model just to view and interact with the bug tracker. Openhub doesn't reflect their mailing lists in *multiple languages*. Appears to be actively maintained mostly by one dev, but notably just got donations supporting him for the next 2–3 years. ++
openldap	libldap-2.4-2	Wide use, vitally important. Very active. Public repo, issue tracker, lists, active development, closing CVEs, the works. This is vitally important, but they appear to have things in hand in terms of project activity. ++
pam	libpam-modules, libpam0g, libpam-modules-bin	Critical central role for authentication. Real site is <a href="http://www.linux-pam.org/">http://www.linux-pam.org/</a> (Debian lists as <a href="http://pam.sourceforge.net/">http://pam.sourceforge.net/</a> , but that points to <a href="https://fedorahosted.org/linux-pam/">https://fedorahosted.org/linux-pam/</a> , which points to this). It has a somewhat active mailing list, and an active issue tracker where issues are clearly opened and closed. Stable version is somewhat old (2013); recent changes in git though relatively small ( <a href="https://git.fedorahosted.org/cgiit/linux-pam.git/">https://git.fedorahosted.org/cgiit/linux-pam.git/</a> ) ++
openssl	libssl1.0.0, openssl	Critical, in wide use. OpenSSL is already funded by CII. ++
net-tools	net-tools	It is unclear whether the Debian package and SourceForge project are still related. This warrants further examination. Upstream listed as <a href="https://developer.berlios.de/projects/net-tools/">https://developer.berlios.de/projects/net-tools/</a> but has moved to <a href="http://sourceforge.net/projects/net-tools/">http://sourceforge.net/projects/net-tools/</a> . Has mailing though rather inactive. Has public git repo, latest commit 2015-01-07 on 2015-03-12. These are tools for controlling the network and typically run as root; it is unclear how exposed they are to network attack (most, if not all, are probably not exposed). This lack of exposure probably lowers their priority. ++
openssh	openssh-client, openssh-server	Vital for security. CII already investing in this. ++
rsyslog	rsyslog	Processes data from untrusted users. Important today; as systemd becomes more used, its use may lessen but still important for merging data sources. Project page <a href="http://www.rsyslog.com/">http://www.rsyslog.com/</a> . Official repo on GitHub, <a href="https://github.com/rsyslog">https://github.com/rsyslog</a> . This is an active multi-person project; in the 1-month period February 12, 2015 through March 12, 2015, excluding merges, 7 authors have pushed 27 commits to master and 40 commits to all branches. On

debian_source	project_name	comment_on_priority
		master, 37 files have changed and there have been 677 additions and 216 deletions. ++
wget	wget	Directly processes potentially malicious data from the Internet, often used in scripts. GNU project. <a href="https://www.gnu.org/software/wget/">https://www.gnu.org/software/wget/</a> . Actual development on GNU's Savannah using git and its bugtracker, <a href="http://savannah.gnu.org/projects/wget/">http://savannah.gnu.org/projects/wget/</a> . Actively developed, many patches. wget-1.16.3 released 2015-03-09 on downloads. Note that this is directly exposed to attack. It has some code documentation (e.g., <a href="http://wget.addictivecode.org/NavigatingTheSource">http://wget.addictivecode.org/NavigatingTheSource</a> ), but it warns that it can be fairly hard to follow, contains a fair number of hacks, and functionality that was "tacked on" (which is not good for security). ++
apr-util	libaprutil1-ldap, libaprutil1	Important. The general Apache Portable Runtime (APR) appears to be actively maintained. The LDAP driver is maintained as part of APR, <a href="http://apr.apache.org/">http://apr.apache.org/</a> . The APR-util version 1.5.4 was released September 22, 2014, so it is actively maintained. They use the apache.org repo and subversion; apr-util has been merged into main trunk, making it a little harder to see LDAP specifics to determine whether the LDAP portion is active. Mailing list <a href="mailto:dev@apr.apache.org">dev@apr.apache.org</a> seems active. ++
coolkey	coolkey	Smart card drivers. Not used in many environments, but critically important for security in other environments. Activity unclear. Managed via Fedora packaging, active here: <a href="https://apps.fedoraproject.org/packages/coolkey">https://apps.fedoraproject.org/packages/coolkey</a> , but that is not where the real code is maintained. Important for users of PKCS#11, but slightly specialized. <a href="http://pkgs.fedoraproject.org/gitweb/?p=coolkey.git">http://pkgs.fedoraproject.org/gitweb/?p=coolkey.git</a> <a href="http://pkg-coolkey.alioth.debian.org/">http://pkg-coolkey.alioth.debian.org/</a> ++
ntp	ntp	Directly connects to network and open to attack. HAS Home Page, NO issue tracker, has mailing list, public repo. CII investing. ++
gnupg	gnupg, gpgv	Vital for protecting email (CII has already invested with a one-time investment). ++
gzip	gzip	Widely-used compression/decompression. Vital, a vulnerability here could be very serious. <a href="http://www.gnu.org/software/gzip/">http://www.gnu.org/software/gzip/</a> Maintained on Savannah ( <a href="http://savannah.gnu.org/projects/gzip/">http://savannah.gnu.org/projects/gzip/</a> ) using git. Last formal release was June 2013. Multiple contributors, but not many. Current git contributions, but only a few a month (2015-02-08, 2015-01-02, 2014-11-10, 2014-10-10), and most seem to be small/trivial (document and syntax tweaks). +++
expat	libexpat1	Parses potentially-dangerous data. This is an XML parser library written in C. Maintenance appears to have effectively halted after its 2012 release. Project at: <a href="http://www.libexpat.org/">http://www.libexpat.org/</a> (It was at, <a href="http://expat.sourceforge.net/">http://expat.sourceforge.net/</a> and its movement is not obvious, so some may be using an older

debian_source	project_name	comment_on_priority
		version; this was follow-on from James Clark's original Expat). On 24 March 2012 Expat 2.1.0 released, it includes security vulnerability fixes for 5 CVEs; Previous release was 2007. "Bug reports" produces error page - tracking does not seem to work. No public mailing list. CVS browsing suggests little or no activity after 2012 (and little before that). Many systems build on top of this library. Note that this doesn't support XML 1.1. +++
freetype	libfreetype6	Dangerous. Processes data from network, including fonts embedded in websites. Main site <a href="http://www.freetype.org/">http://www.freetype.org/</a> Last release FreeType 2.5.5, 2014-12-30. Git repo. Very active git repo by multiple contributors. ++
libgcrypt11	libgcrypt11	Crypto library, security-relevant. LGPL crypto library. Part of GnuPG (GNU Privacy Guard). <a href="http://ftp.gnupg.org/gcrypt/libgcrypt/">http://ftp.gnupg.org/gcrypt/libgcrypt/</a> . Already selected by CII. ++
keyutils	libkeyutils1	Manages authorization and encryption keys required to perform secure operations. Public Repo, NO bug tracker, NO mailing list. ++
xz-utils	liblzma5, xz-utils	Widely-used compression/decompression. Vital, a vulnerability here could be very serious. Has Home page, public repo, Fairly active ML <a href="http://www.mail-archive.com/xz-devel@tukaani.org/">http://www.mail-archive.com/xz-devel@tukaani.org/</a> , NO bug tracker ++
p11-kit	libp11-kit0	Manages security keys. Has Home page, Bug Tracker <a href="https://bugs.freedesktop.org/enter_bug.cgi?product=p11-glue">https://bugs.freedesktop.org/enter_bug.cgi?product=p11-glue</a> , ML <a href="http://lists.freedesktop.org/archives/p11-glue/">http://lists.freedesktop.org/archives/p11-glue/</a> fairly active ++
pcre3	libpcre3	Regexes are widely used for security-relevant input validation. Has Home page, Bug Tracker <a href="http://bugs.exim.org/buglist.cgi?product=PCRE">http://bugs.exim.org/buglist.cgi?product=PCRE</a> NOT very active, public repo <a href="http://sourceforge.net/projects/pcre/files/">http://sourceforge.net/projects/pcre/files/</a> ++
cyrus-sasl2	libsasl2-2, libsasl2-modules	Authentication library. Has Home Page, HAS bug tracker <a href="https://bugzilla.cyrusimap.org/index.cgi">https://bugzilla.cyrusimap.org/index.cgi</a> , fairly active ML <a href="http://asg.andrew.cmu.edu/archive/index.php?mailbox=archive.cyrus-sasl">http://asg.andrew.cmu.edu/archive/index.php?mailbox=archive.cyrus-sasl</a> ++
libxml2	libxml2	Processes attacker-provided data. Has Home Page, Has bug tracker <a href="https://bugzilla.gnome.org/buglist.cgi?product=libxml2">https://bugzilla.gnome.org/buglist.cgi?product=libxml2</a> which is very active. ++
shadow	login	Front door for local login. Has Home Page, part of shadow tool suite, Latest release May 2014, Public Repo, Somewhat active mailing list, NO issue tracker. ++
tar	tar	Widely-used compression/decompression. Vital, a vulnerability here could be very serious. Has Home page, mailing list, Public Repo <a href="git://git.savannah.gnu.org/tar.git">git://git.savannah.gnu.org/tar.git</a> , uses "technical support" for issue tracking (could be confusing). Does have changes in git repo, but only a few a month and typically small. ++

debian_source	project_name	comment_on_priority
zlib	zlib1g	Widely-used compression/decompression. Vital, a vulnerability here could be very serious. Has Home Page, has mailing list <a href="mailto:zlib@gzip.org">zlib@gzip.org</a> , NO issue tracker, Latest release April 2013 (old). +++
apr	libapr1	Important. The general Apache Portable Runtime (APR) appears to be actively maintained. However, it's not as clear that the LDAP library in it is as actively managed. Has Home Page, HAS Mailing List, Issue tracker <a href="https://bz.apache.org/bugzilla/enter_bug.cgi?product=APR">https://bz.apache.org/bugzilla/enter_bug.cgi?product=APR</a> ++
libjpeg8	libjpeg8	Widely-used format that is decompressed and provided by attackers. HAS Home Page, Can download as tarball, NO issue tracker, latest version/activity Jan 2014, NO mailing list +++
libpng	libpng12-0	Widely-used format that is decompressed and provided by attackers. HAS Home Page, Can Access Source code, Not much activity - just warnings about CVEs that were fixed. +++
libressl	libressl	Fork of OpenSSL. Has Home Page, Can Download as tarball, VERY Active on Github <a href="https://github.com/libressl-portable/portable">https://github.com/libressl-portable/portable</a> ++
unzip	unzip	Widely-used format that is decompressed and provided by attackers. Has Home Page, Can Download Binaries, NO issue tracker. +++
giflib	libgif4	Widely-used format that is decompressed and provided by attackers. Has Home Page, Active issue tracker, Mailing list, public repo. ++
mod-gnutls	libapache2-mod-gnutls	This is in widespread use for SSL/TLS on web servers; it uses GnuTLS instead of OpenSSL. Old website <a href="http://www.outoforder.cc/projects/apache/mod_gnutls/">http://www.outoforder.cc/projects/apache/mod_gnutls/</a> points to newer site " <a href="https://mod.gnutls.org/?outoforder-ref">https://mod.gnutls.org/?outoforder-ref</a> ". At newer site the mailing list has a few relevant postings/month, though the March 2015 postings raise concerns about dropped patches. The bug tracker is badly spammed and one seems to be fixing/addressing it. Last release 17-Feb-2014. This looks like it badly needs help. +++
postfix	postfix	"This is a widely-used mail transfer agent (MTA). Postfix was created by Wietse Venema, main site <a href="http://www.postfix.org">www.postfix.org</a> . On February 8, 2015 Postfix released 3.0.0, so currently active. Developer mailing list postfix-devel has activity. (e.g., <a href="http://news.gmane.org/gmane.mail.postfix.devel/cutoff=2948">http://news.gmane.org/gmane.mail.postfix.devel/cutoff=2948</a> ). There is no public repo; Wietse (the lead developer) does not want to allow any external review before formal release, as noted in <a href="https://groups.google.com/forum/#!topic/mailling.postfix.users/6Kkel3J_nv4">https://groups.google.com/forum/#!topic/mailling.postfix.users/6Kkel3J_nv4</a> (The Debian package may be an obsolete fork; the Debian package has been assembled by LaMont Jones from sources available from <a href="http://www.postfix.org">http://www.postfix.org</a> , and can be

debian_source	project_name	comment_on_priority
		cloned from git via: git clone git://git.debian.org/~lamont/postfix.git. Changes on <a href="http://anonscm.debian.org/cgiit/users/lamont/postfix.git/log/">http://anonscm.debian.org/cgiit/users/lamont/postfix.git/log/</a> in git stop in 2011.). ++"
cryptsetup	cryptsetup	This is used to set up disk encryption (at rest) (e.g., Linux Unified Key Setup (LUKS)). Home page <a href="http://code.google.com/p/cryptsetup">http://code.google.com/p/cryptsetup</a> , uses git. Tracker in active use. Latest version 1.6.6 released 2014-08-16. Development ongoing; 8 commits in February 2015. ++

### C. Potential Improvements to Other Sites to Improve Data

Some changes to other sites would make tasks like this much easier:

1. Debian could include in their packages detailed automatically extractable URLs that identify their upstream source (and if different, the repository they directly use). This information is often in the “copyright” information, but not necessarily in a specific format, and it is often stale (e.g., if the project moves, the Debian data often includes only the old URL).
2. Individual projects should have publicly visible version control repositories, discussion mechanisms (e.g., mailing list), and issue trackers (security-relevant issues may be temporarily private). Projects can use GitHub, SourceForge, Savannah, and similar systems to easily meet these needs; if they choose to self-host, they should do so in a standardized way so that automated crawlers can gather data across multiple projects. Doing this makes it much easier for third parties to determine whether a project is active. There are good reasons to do this anyway; potential users and developers will also want to see this, to give them confidence that the software is being actively maintained by multiple people.
3. Issue trackers should have a clear and standard mechanism for distinguishing bugs from enhancement requests. Bug response times are important indicators for project responsiveness, and mixing them up with enhancement requests makes the data less useful. It would be very helpful if sites like GitHub would build in, by default, some tags for this purpose when projects start.

## 7. Conclusions

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Some OSS projects are more concerning from a security point of view than others. We have used quantitative analysis, combined with human knowledge, to identify especially concerning OSS projects for further (human) investigation and possible investment. It would be a mistake to assume that all OSS is insecure. It would also be a mistake to assume that just because software has an OSS license it must be secure. Instead, OSS should be evaluated; if a particular OSS project is important but insecure, steps should be taken to improve or replace it.

This work does not “turn the crank and produce the answers.” There are many reasons for this; for example, limited metrics were available in the limited time we had, and some metrics, such as the metrics for bug report responses, turned out to be difficult to acquire in the time we had. Measuring software security is notoriously difficult, so we did not attempt to do it directly. The data sources we had were “messy”; for example, it is difficult to map names to the correct project websites, and data such as number of contributors can be difficult to get automatically across many projects. Indeed, project websites may not give a full picture of the actual situation.

An additional challenge is that a lot of human knowledge is not easily captured. For example, “ftp” (the client) and “netcat-traditional” rank as high risk; that makes sense because they definitely interact with the network and thus create opportunities for remote attack. What is more, the ones used by Debian do not appear to be very active projects. Are they *really* that critical, though? A vulnerability in netcat, while undesirable, is unlikely to have the same impact as a vulnerability in openssl, openssh, or some other programs.

There are also some strategic questions in terms of what kinds of programs should receive investment. For example, Mail Transfer Agents (MTAs) scored higher than many other programs in terms of risk, in part because they are directly exposed to attack from the network. Should security investments be made in MTAs, or would the money be better spent elsewhere?

Thus, though we provide a list of programs identified solely by our metrics, we also provide a list of programs that were identified as higher risk, and then selected the most likely ones from them. That is the process we would recommend to others as well. That said, we were guided by the data, especially for identifying the larger set to then be examined further. We have provided our initial cut using this process. We believe the next step is to further investigate these OSS projects for security and project healthiness. The

more detailed “results.csv” file includes the complete list of packages and data that we have gathered about them.



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## Acronyms and Abbreviations

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*BSD	One of the descendants of the BSD operating system, such as FreeBSD, NetBSD, and OpenBSD
AOSP	Android Open Source Project
API	Application Programming Interface
AUC	Area Under the ROC Curve
BFD	Binary File Descriptor
BSD	Berkeley Software Distribution
BY	Attribution (a Creative Commons license)
CC	Creative Commons
CCCC	C and C++ Code Counter
CII	Core Infrastructure Initiative
CMS	Content Management System
CVE	Common Vulnerabilities and Exposures
CVSS	Common Vulnerability Scoring System
DEA	Data Envelopment Analysis
DHS	Department of Homeland Security
FAQ	Frequently Asked Questions
FIPS	Federal Information Processing Standards
FLOSS	Free/Libre/Open Source Software
FS	Free Software
FSF	Free Software Foundation
FTP	File Transfer Protocol
GCC	GNU Compiler Collection
GNU	GNU's Not Unix
GPG	GNU Privacy Guard
GPL	(GNU) General Public License
GTRI	Georgia Tech Research Institute
HOST	Homeland Open Security Technology
IDA	Institute for Defense Analyses
IMAP	Internet Message Access Protocol
IRC	Internet Relay Chat
KSLOC	Thousand Source Lines of Code
KVM	Kernel-based Virtual Machine
LF	Linux Foundation
LR	Logistic Regression
MIT	Massachusetts Institute of Technology
MTA	Mail Transfer Agent
NIST	National Institute of Standards and Technology
NTP	Network Time Protocol
NVD	(NIST) National Vulnerability Database

OS	Operating System
OSI	Open Source Initiative
OSMM	Open Source Maturity Model (note: several different projects use this name)
OSS	Open Source Software
OWASP	Open Web Application Security Project
PAM	Pluggable Authentication Module
PHP	PHP: Hypertext Preprocessor
QSOS	Qualification and Selection of Open Source software
RDMA	Remote Direct Memory Access
REST	Representational State Transfer
ROC	Receiver Operating Characteristic
RPC	Remote Procedure Call
SLOC	Source Lines of Code
SSL	Secure Sockets Layer
SVM	Support Vector Machines; see [Mishra2012]
TLS	Transport Layer Security
Y-O-Y	Year-Over-Year

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14. ABSTRACT Some open source software (OSS) is widely used and depended on, and yet not received the level of security analysis appropriate to its importance. This paper describes our work to help identify OSS projects that may especially need investment for security by identifying and using metrics. We performed a literature search, identified promising metrics and potentially-concerning software packages to investigate, developed a specific approach, and applied it to identify a set of OSS projects that we believe are especially concerning. We have focused on automatically gathering metrics, especially those that suggest less active projects. For our initial set of projects to examine we took the set of software packages installed by Debian base and added packages that we or others identified as potentially concerning; we could easily add more projects to consider in the future.					
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